The Stricter Standard: An Empirical Assessment of Daubert’s Effect on Civil Defendants

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The Stricter Standard: An Empirical Assessment of Daubert’s Effect on Civil Defendants

Cover Page Footnote
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THE STRICHER STANDARD: AN EMPIRICAL ASSESSMENT OF DAUBERT’S EFFECT ON CIVIL DEFENDANTS

Andrew Jurs*
Scott DeVito++

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The question of whether Daubert v. Merrell Dow Pharmaceuticals, Inc.\textsuperscript{1} adopted a more lenient or more stringent standard for testing the reliability of expert evidence has dogged academics, practitioners, and researchers for twenty years.\textsuperscript{2} Research since 1993, using a variety of methodologies, has been largely inconsistent. Some quantitative studies show that the Daubert standard has had no effect on the admission of expert evidence.\textsuperscript{3} Other quantitative studies find the opposite, that Daubert is a more stringent standard of admissibility.\textsuperscript{4} Daubert cannot be both meaningless and more restrictive, and determining which of these two conclusions is correct will have enormous implications.

If the shift to Daubert had no effect on the judicial management of evidence or trial outcomes, the choice between the Daubert standard and the traditional Frye test for admissibility\textsuperscript{5} is without consequence.\textsuperscript{6} Perhaps, if this is true, reformers should abandon efforts to change the tort system through doctrinal modifications and seek redress elsewhere.\textsuperscript{7} Moreover, an impotent Daubert

\begin{itemize}
\item[2.] See Erica Beecher-Monas, Blinded by Science: How Judges Avoid the Science in Scientific Evidence, 71 TEMP. L. REV. 55, 75–76 (1998) (indicating that the issue of whether Daubert imposes a stricter standard for the admissibility of scientific evidence is a recurring topic of debate); see also Sophia I. Gatowski et al., Asking the Gatekeepers: A National Survey of Judges on Judging Experience in a Post-Daubert World, 25 LAW & HUM. BEHAV. 433, 443 (2001) (concluding that judges are split in their views on the effect Daubert has had in their analyses of whether to admit expert evidence).
\item[4.] See Lloyd Dixon & Brian Gill, Changes in the Standards for Admitting Expert Evidence in Federal Civil Cases Since the Daubert Decision xv (2001) (“Our analysis of district court opinions suggests that after Daubert, judges scrutinized reliability more carefully and applied stricter standards in deciding whether to admit expert evidence.”); see also Carol Krafka et al., Judge and Attorney Experiences, Practices, and Concerns Regarding Expert Testimony in Federal Civil Trials, 8 PSYCHOL. PUB. POL’Y & L. 309, 322 (2002) (noting that judges “reported that they were more likely to scrutinize expert testimony before trial and were less likely to admit it” following Daubert).
\item[5.] Frye v. United States, 293 F. 1013, 1014 (D.C. Cir. 1923) (holding that scientific evidence must have gained sufficient “standing and scientific recognition among” authorities in the field).
\item[7.] See id. at 505 (arguing that procedural limits on tort litigation often lead to tort reform movements that have no effect on the outcomes of subsequent cases).
\end{itemize}
would also indicate that the controversy in evidence law about expert admissibility has been much ado about nothing.

However, if *Daubert* is a more stringent standard, the implications are enormous. It would indicate that—at the time of its decision—the Supreme Court badly misjudged what effect *Daubert* would have on the admissibility of scientific evidence. At the time, the Court believed that the decision would be consistent with "the liberal thrust of the Federal Rules [of Evidence] and their general approach of relaxing the traditional barriers to opinion testimony." If this were not true, the Court would need to revisit expert reliability standards, to clarify the standard by either affirming the new reality or insisting upon the lenient standard initially envisioned. A stricter *Daubert* standard would also have an enormous effect on substantive tort law by procedural modification. Finally, a stricter standard for admissibility affects the market for expert witness services, because only those experts who can provide a sound empirical basis for their opinions will be permitted to testify. Therefore, the interpretation of the *Daubert* standard is important, especially in an era when expert witness testimony is as prevalent as it is today.

In this Article, we answer the question: Did *Daubert* have a measurable effect on expert admissibility, and if so, did it adopt a stricter or more lenient standard for admissibility? To make this determination, this Article builds on a


10. Michael J. Saks, *The Aftermath of Daubert: An Evolving Jurisprudence of Expert Evidence*, 40 JURIMETRICS J. 229, 239 (2000) ("Testimony from those fields that could, but do not, provide adequate data . . . should be excluded until they can provide adequate empirical support for what the experts are claiming for their field in general, or for themselves individually.").

11. The percentage of cases involving experts remains an area with surprisingly few definitive statistics. One oft-cited study concluded that experts are involved in as many as 86% of all civil jury cases. Samuel R. Gross, *Expert Evidence*, 1991 WISC. L. REV. 1113, 1119 (reporting data from all civil jury cases tried in California Superior Court between 1985 and 1986). Other studies returned an expert witness rate between 63% and 71%. Anthony Champagne et al., *An Empirical Examination of the Use of Expert Witnesses in American Courts*, 31 JURIMETRICS J. 375, 380 (1991) (finding that 57 out of 90 civil cases tried in the Dallas County District Court in Texas in 1988 utilized expert witness testimony); Daniel Shuman et al., *An Empirical Examination of the Use of Expert Witnesses in the Courts—Part II: A Three City Study*, 34 JURIMETRICS J. 193, 197 (1994) (reporting that 131 cases out of 183 included expert testimony, a rate of 73%). Regardless of the exact rate, experts are a permanent and enormous part of modern litigation.
methodology used in a 2005 study by Edward Cheng and Albert Yoon. In certain cases, removal allows a civil defendant the option of transferring a case from state to federal court. By removing the case, however, the defendant may also be changing the scientific admissibility standard from the state court’s *Frye* standard to the federal *Daubert* standard. Cheng and Yoon first suggested that the effect of *Daubert* could be measured across the judicial system by reviewing removal rates from state to federal court in multiple jurisdictions. We also believe that removal rates offer a very accurate method of measuring the systematic effect of *Daubert* because these rates test litigants’ opinions in the context of real cases.

To this end, we created a database of approximately 4 million cases and calculated removal rates during the period from 1990 to 2000. Based on our two-step analysis of that data, we conclude that *Daubert* is a stricter standard than *Frye* for the admissibility of expert testimony. First, by properly identifying, isolating, and removing other possible confounding variables, we were able to isolate and then measure the effect of the federal courts’ adoption

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13. See 28 U.S.C. § 1446 (2006) (articulating the requirements to remove a case from state court to federal court); see also Cheng & Yoon, *supra* note 6, at 482–83 (explaining that a litigant may remove a tort action—normally restricted to state court—to federal court under diversity jurisdiction).

14. The central premise allowing for this comparison is that evidence rules are generally considered to be procedural, and thus a state’s evidentiary rules on expert admissibility will not transfer upon removal. Instead, the Federal Rules of Evidence will usually apply to the action, even if state substantive tort law remains. See Cheng & Yoon, *supra* note 6, at 483 (explaining that, although state substantive law is retained in tort cases because of diversity jurisdiction and the *Erie* doctrine, the procedural rules of the new forum are applied (citing Erie R.R. Co. v. Tompkins, 304 U.S. 64, 78 (1938))); see also, e.g., Robin Kundis Craig, *When Daubert Gets Erie: Medical Certainty and Medical Expert Testimony in Federal Court*, 77 DENV. U. L. REV. 69, 82 (1999) (asserting that “[i]n most federal courts apply the Federal Rules of Evidence rather than state rules to all evidentiary questions in diversity cases”); Abbe R. Gluck, *Intersystemic Statutory Interpretation: Methodology as “Law” and the *Erie* Doctrine*, 120 YALE L.J. 1898, 1979 & n.279 (2011) (explaining that the Federal Rules of Evidence are generally considered to be procedural and are applied in cases removed from state courts); Alex Stein, *Constitutional Evidence Law*, 61 VAND. L. REV. 65, 98–99 (2008) (noting that all evidence rules, except burdens of proof, privileges, presumptions, and the competency of witnesses, are procedural).

15. See Cheng & Yoon, *supra* note 6, at 482 (using the rate at which cases were removed to federal court to measure the effect of the *Daubert* decision on the admission of expert evidence).

16. See *infra* Part II.B (explaining the merits of using removal rate as a measuring tool).

17. See *infra* Part II.C.2 (demonstrating that the rate of removal to federal court in *Frye* states increased following the adoption of *Daubert* in federal courts); see also *infra* Part II.C.3 (demonstrating that the rate of removal to federal court decreased once *Frye* states adopted *Daubert*); *infra* Part II.C.4 (concluding that the rates of removal to federal court in relation to a jurisdiction’s adoption of *Daubert* indicate that *Daubert* is a stricter standard for admissibility than *Frye*).
of *Daubert* on removal rates to federal court.18 These results demonstrate that, absent other countervailing effects, the adoption of *Daubert* by federal courts results in an increase in civil defendants removing the case to federal courts to benefit from the courts’ restriction of expert testimony under that standard.19 Second, by using the same process, we were able to measure the counter-effect on removals when a state court adopted *Daubert* after the federal courts had already done so.20 These results demonstrate that a state court’s adoption of *Daubert* after federal adoption decreases the rate of removals to federal court.21 This is consistent with our initial result because, after the state court adopts *Daubert*, litigants gain no procedural benefit from shifting court systems when both apply the same reliability standard. Combined, these results demonstrate unequivocally that, when measured in the aggregate and based on actual behavior in real cases, civil defendants believe the *Daubert* standard is more restrictive to expert testimony and act accordingly.22

We begin in Part I of this Article by briefly outlining the *Daubert* standard and the impact the case has had on gatekeeping expert testimony. Additionally, Part I introduces prior empirical studies on *Daubert*, and addresses those studies’ contradictory results. We then begin our empirical analysis in Part II. First, we briefly review the Cheng and Yoon study and its counterintuitive finding that a state’s choice between *Daubert* and *Frye* “makes no difference in practice.”23 We then explain in more detail why the removal rate metric is the key to determining *Daubert*’s true effect. Next, we then used the removal rate metric to analyze our database of civil case data in order to measure the aggregate effect of *Daubert* on litigant behavior. Our analysis demonstrates that the adoption of *Daubert* in federal court did, in a statistically significant manner, increase removal rates from *Frye* states. Furthermore, we will also demonstrate that in the period after the federal adoption of *Daubert*, a state’s adoption of a *Daubert*-type standard for state court reduced the removal rates, thereby reversing the post-*Daubert* increase in removals. Both findings demonstrate, for the first time and using actual case data, that civil defendants believe *Daubert* is a stricter admissibility standard.

18. See infra Part II.C.2 (explaining the methodology by which we analyzed the impact of *Daubert* on removal rates in *Frye* states).
19. See infra Part II.C.2 (noting our conclusion that the rate of removal to federal court in *Frye* states increased after the adoption of *Daubert* in federal courts).
20. See infra Part II.C.3 (explaining the methodology by which we analyzed the subsequent adoption of *Daubert* in state courts on removal rates).
21. See infra Part II.C.4 (noting our conclusion that the rate of removal to federal court following a state’s adoption of *Daubert* reverted to pre-federal-court-adoptation levels).
22. See infra Part II.C.4 (summarizing our conclusion that *Daubert* is a stricter standard because of the preceding analysis of litigant behavior in relation to a jurisdiction’s adoption of that standard).
23. Cheng & Yoon, supra note 6, at 503.
Because our results so clearly contrast with Cheng and Yoon, who used the same data, we then revisit the Cheng and Yoon study and evaluate its methodology in detail. In so doing, we found that Cheng and Yoon appear to have made three kinds of errors in their analysis that explain why they failed to identify the true nature of adoption of Daubert. First, there seem to be errors in how they determined the removal rates that form their study population. Second, they failed to recognize that state adoption of Daubert after federal adoption of Daubert will have a different effect than state adoption of Daubert before federal adoption of Daubert. Consequently, their data is internally inconsistent with regard to their analysis of state adoption of Daubert. Third, the small size of the study population they directly use (removal rates) combined with the enormous size of the population they derive the study population from (torts filed in state court) minimized the chances of finding an effect from state adoption of Daubert.

Finally, we will finish in Part III with some comments on the significance of a stricter Daubert standard. We also suggest areas for future empirical consideration that would build upon our research.

By measuring the actual behavior of civil litigants, this Study can answer the question whether Daubert is a stricter standard for expert admissibility. Based on our analysis of the behavior of civil defendants in actual cases, the answer is yes.

I. GATEKEEPING EXPERT EVIDENCE, DAUBERT, AND CURRENT RESEARCH ON DAUBERT’S EFFECT

As many as 86% of all cases involve expert evidence and, consequently, this evidence is of central importance to modern civil and criminal litigation.24 With certain civil causes of action—such as products liability or medical malpractice—cases rely on expert testimony; see also Joseph Sanders, From Science to Evidence: The Testimony on Causation in the Benedictin Cases, 46 STAN. L. REV. 1, 31–32 (1993) (noting the particular importance of expert witnesses in products liability cases).

Before analyzing the data on Daubert’s effect, we will first review the issue of admissibility

24. See supra note 11 and accompanying text (discussing the prevalence of expert evidence in modern litigation).

25. See Gross, supra note 11, at 1119 (finding that over 90% of products liability and medical malpractice cases relied on expert testimony); see also Joseph Sanders, From Science to Evidence: The Testimony on Causation in the Benedictin Cases, 46 STAN. L. REV. 1, 31–32 (1993) (noting the particular importance of expert witnesses in products liability cases).

26. See Krafta et al., supra note 4, at 328 (reporting that approximately 69% of judges surveyed changed their procedures for considering the admissibility of expert evidence following their jurisdictions’ shift to Daubert and that, generally, judges view themselves as more active in the admissibility process following Daubert); see also The Changing Role of Judges in the Admissibility of Expert Evidence, CIV. ACTION (Nat’l Ctr. for State Courts), Spring 2006, at 3 [hereinafter Changing Role of Judges].
standards for scientific evidence, and review prior interesting but inconclusive studies on their effect.

A. The Frye Decision and Expert Admissibility

Concern over the admissibility of expert witness testimony did not originate with the Daubert debate of the late twentieth century. In fact, the admissibility standard articulated in Frye v. United States grew out of dissatisfaction with preexisting common law standards. Eventually the Frye standard became the admissibility test in federal and many state courts, when litigants raised the issue of admissibility of scientific testimony. Frye required that—before admission—a method be generally accepted in the relevant field.27 However, throughout the twentieth century, critics exposed many weaknesses of the Frye general acceptance formula.28

Before 1923, judges used two common law tests to determine the admissibility of expert evidence. Some courts simply evaluated the helpfulness of the evidence to a lay jury, and admitted the evidence if it was relevant.29 Others would consider the qualifications of the expert under what Professor David Faigman and his coauthors call “the commercial marketplace test,” whereby any expert who succeeds in a chosen profession must do so based on expertise.30

By 1923, the problems with relying on relevance or the reputation of the expert led the D.C. Circuit to reconsider the standard for admissibility of expert evidence.31 In Frye, the court shifted the focus away from the prior

27. Frye v. United States, 293 F. 1013, 1014 (D.C. Cir. 1923).
28. See infra notes 36–41 and accompanying text (tracking the evolution of the Frye standard and its shortcomings).
29. Kaye et al., supra note 3, § 1.2.1, at 6–7 (noting that expert testimony must have been “particularly important to aiding the trier of fact” in order to be admitted); David L. Faigman et al., Check Your Crystal Ball at the Courthouse Door, Please: Exploring the Past, Understanding the Present, and Worrying About the Future of Scientific Evidence, 15 CARDOZO L. REV. 1799, 1803 (1994) (explaining that the relevant inquiry was whether the testimony was from an area beyond the knowledge of the average juror).
30. Faigman et al., supra note 29, at 1804 & n.13 (noting that judges would evaluate the qualifications and expertise of the expert through “the expert’s success in an occupation or profession which embraced” the subject matter in question); Michael J. Saks, Judging Admissibility, 35 J. CORP. L. 135, 136 (2009) (explaining that judges often inferred expertise from the expert’s commercial success).
31. Frye, 293 F. at 1013; see also Faigman et al., supra note 29, at 1805 (arguing that Frye was a response to the failure of the commercial marketplace test); Saks, supra note 30, at 137 (noting that “commercial value is not a measure of scientific or any other kind of validity”); Michael J. Saks, Merlin and Solomon: Lessons from the Law’s Formative Encounters with Forensic Identification Science, 49 HASTINGS L.J. 1069, 1074 (1998) (explaining that the flaws of the commercial marketplace test necessitated the Frye standard).
common law standard to a “general acceptance test.”\textsuperscript{32} The court concluded that expert testimony cannot be considered at trial unless it has become “sufficiently established as to have gained general acceptance in the particular field to which it belongs.”\textsuperscript{33} Applying this standard, the \textit{Frye} court affirmed the exclusion of the expert testimony because the testing in question remained experimental.\textsuperscript{34}

With the holding in \textit{Frye}, the D.C. Circuit developed a standard by which a court could assess whether a scientific principle had gained sufficient recognition as to be appropriate to consider in court. Yet, for all its controversy, particularly later, the standard appears to have been applied sparingly for decades. Professor Michael Saks notes that the case was not cited for over 10 years following the decision; it received fewer than 15 citations in its first 25 years, and fewer than 100 citations in the next 25-year span.\textsuperscript{35} By the late twentieth century, however, the trend shifted.

The \textit{Frye} test was used primarily in criminal cases until the late 1980s\textsuperscript{36} when courts began to apply it to complex scientific causation in toxic tort cases.\textsuperscript{37} The application of the \textit{Frye} test to complicated civil cases intensified the existing controversy over the standard, and commentators noted the unworkability of the vague “general acceptance” test.\textsuperscript{38} Critics argued that the

\begin{footnotesize}
\textsuperscript{32} Frye, 293 F. at 1013–14 (considering the admissibility of expert testimony explaining the systolic blood pressure deception test); see also Faigman et al., \textit{ supra} note 29, at 1806 (arguing that, although \textit{Frye} is inherently a marketplace test, the standard shifts the emphasis from the credentials of the expert and the commercial marketplace, requiring an evaluation of the body of knowledge rather than the expert himself).

\textsuperscript{33} Id., 293 F. at 1014.

\textsuperscript{34} Id. (holding that “the systolic blood pressure deception test had not yet gained such standing and scientific recognition” to justify admitting the expert testimony).

\textsuperscript{35} Saks, \textit{ supra} note 30, at 139; see also Faigman et al., \textit{ supra} note 29, at 1808 (describing \textit{Frye} as “barely noticed” and “of little importance” in the decades after it was decided). Professor David Bernstein noted that, despite the dearth of federal court citations to \textit{Frye}, state courts may have used the standard more often in deciding cases. David E. Bernstein, \textit{Frye, Again: The Past, Present, and Future of the General Acceptance Test}, 41 \textit{JURIMETRICS J.} 385, 388–90 (2001). However, that theory cannot be tested because state trial court opinions generally remain unpublished. \textit{Id}.

\textsuperscript{36} Bernstein, \textit{ supra} note 35, at 390 (asserting that \textit{Frye} was applied only in criminal cases until 1988); Kenneth J. Chesebro, \textit{Galileo’s Retort: Peter Huber’s Junk Scholarship}, 42 \textit{AM. U. L. REV.} 1637, 1693–95 (1993) (indicating that 64 of 67 federal appellate decisions applying \textit{Frye} were criminal decisions); Susan Haack, \textit{An Epistemologist in the Bramble-Bush: At the Supreme Court with Mr. Joiner}, 26 \textit{J. HEALTH POL’Y & L.} 217, 227 (2001) (noting that \textit{Frye} was the standard for the admissibility of expert evidence first in criminal cases, but later became the standard in civil cases); \textit{cf.} KAYE ET AL., \textit{ supra} note 3, \S 5.3.2, at 159–60 (detailing the several types of evidence the \textit{Frye} test has been used to exclude).

\textsuperscript{37} Bernstein, \textit{ supra} note 35, at 390–92 (explaining that \textit{Frye} was applied to toxic tort cases starting in 1988).

\textsuperscript{38} \textit{See}, e.g., Bert Black, \textit{A Unified Theory of Scientific Evidence}, 56 \textit{FORDHAM L. REV.} 595, 628 (1988) (arguing that the \textit{Frye} test is too incoherent to be applied effectively by the
test provided little guidance on what evidence demonstrated general acceptance, and then what level of proof was required.\footnote{39} They also noted Frye’s failure to exclude “scientific” evidence that later proved to be unfounded.\footnote{40} In his exhaustive study of toxic tort litigation, Professor Joseph Sanders analyzed the judicial decision making process in case law and examined in detail how the system failed to properly manage complex expert testimony.\footnote{41}

Criticism of the Frye standard left many with an impression of judicial incompetence in the management of scientific evidence,\footnote{42} and commentators showed increasing dismay with courts’ inability to sift away “junk science.”\footnote{43} In the context of this debate over the courts’ ability to properly manage complex science, the U.S. Supreme Court granted certiorari in a toxic tort case from the Ninth Circuit: Daubert v. Merrell Dow Pharmaceuticals, Inc.\footnote{44}

\begin{verbatim}
39. See KAYE ET AL., supra note 3, § 5.3.3, at 164 (noting that the Frye decision was unclear in its standard); Giannelli, supra note 41, at 1219 (finding that courts are unsure as to the appropriate scenarios to which to apply the Frye standard).
40. See PETER W. HUBER, GALILEO’S REVENGE (1991); see also KAYE ET AL., supra note 3, § 5.3.2, at 163–64 (providing examples of cases in which the Frye standard allowed for the admission of “worthless and potentially misleading” scientific evidence); Giannelli, supra note 38, at 1224–25 (pointing to the admission of evidence collected from paraffin tests as proof of Frye’s shortcomings).
41. Sanders, supra note 25, at 27–28 (arguing that scientific data “fails to provide the lay fact finder with the resources necessary to assess properly the quality of experts or the weight and relative importance of the scientific findings”).
42. See Andre A. Moenssens, Admissibility of Scientific Evidence—An Alternative to the Frye Rule, 25 WM. & MARY L. REV. 545, 551–52 (1984) (asserting that judges often admitted evidence under Frye because of their failure to understand the relevant scientific principles).
\end{verbatim}
B. The Daubert Revolution and Its Aftermath

The Federal Rules of Evidence became effective in 1975, and while the rules contained a specific section on expert admission, the effect of Rule 702 on the Frye standard was unclear. The Daubert Court definitively resolved the dispute.

In Daubert, Justice Harold Blackmun—writing for the majority—stated unequivocally “[t]hat [the] austere [Frye] standard, absent from, and incompatible with, the Federal Rules of Evidence, should not be applied in federal trials.” Beyond that initial labeling of Frye as austere, there are additional reasons to believe that the Daubert Court consciously lowered the standard for admissibility. Instead of Frye’s “general acceptance” test, the Daubert test depends on the twin standards of Rule 702: relevance and reliability. The Court specifically stated that a Rule 702 approach is consistent with the Federal Rules’ purpose “of relaxing the traditional barriers to opinion testimony.” In addition, under the new standard, the Court specifically endorsed the idea that shaky expert evidence should be admitted, subject to vigorous cross-examination. Four years later in General Electric Co. v. Joiner, the Court expressly stated what had been implied in Daubert: “the Federal Rules of Evidence allow district courts to admit a somewhat broader range of scientific testimony than would have been admissible under Frye.” The intent of the Daubert standard, the Court later explained, was to ensure that any expert “employs in the courtroom the same level of intellectual rigor that characterizes the practice of an expert in the relevant field.”


46. See Giannelli, supra note 38, at 1228–29 (“The adoption of the Federal Rules of Evidence has not resolved the uncertain status of the Frye test. Indeed, the Federal Rules, which became effective in 1975 and have been adopted in various forms by twenty-two jurisdictions, have contributed to the confusion.” (citation omitted)); see also Faigman et al., supra note 29, at 1809–10 (noting that, even though Rule 702 and the Advisory Committee Notes made no reference to Frye, courts incorporated the “general acceptance” standard into the Rule 702 test).


48. See id. at 589 (explaining that Rule 702 is “[t]he primary locus” of the trial judge’s duty to ensure that expert evidence is both relevant and reliable; see also FED. R. EVID. 702 (requiring expert evidence to be both helpful to the trier of fact and reliable in both its methodology and the expert’s execution of that methodology).

49. Id. at 588 (internal quotation marks omitted) (citing Beech Aircraft Corp. v. Rainey, 488 U.S. 153, 169 (1988)).

50. Id. at 596 (citing Rock v. Arkansas, 483 U.S. 44, 61 (1987)) (explaining that inconsistencies or inaccuracies in expert testimony can be challenged by cross-examination).


After *Daubert*, federal courts began to apply the Rule 702 standard to many different forms of scientific evidence. Daubert empowered judges to act as gatekeepers of expert evidence, a role that requires detailed assessment of complex scientific principles. Even though the Daubert majority expressed confidence in the judiciary’s ability to handle this gatekeeping role, Chief Justice William Rehnquist expressed apprehension in converting judges into “amateur scientists.”

Certainly many commentators, scholars, and even federal judges shared his concern.

After 20 years, *Daubert’s* influence in state courts remains mixed. Many states decided, even after *Daubert*, that their courts would continue to use the *Frye* test to evaluate expert evidence. Other states concluded that *Daubert* recognized the judicial role already present in their jurisdictions. Finally, a large group of states followed the lead of the Supreme Court and abandoned *Frye* in favor of *Daubert’s* rules-based reliability approach. In these states, the *Daubert* decision directed judges to become more involved in gatekeeping decisions and screening scientific evidence for methodological soundness.

Indeed, even in those states that retained *Frye*, the debate over the

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53. See generally Saks, supra note 30, at 147–55 (discussing the application of the *Daubert* standard to several types of forensic evidence).
54. *Daubert*, 509 U.S. at 592–94, 597 (requiring the trial judge to assess the scientific validity of proffered expert evidence and urging judges to engage in the evaluation of such factors as the peer reception and margin of error of the scientific technique or methodology in question); see also Gatowski et al., supra note 2, at 436–37 (discussing the factors *Daubert* provided to judges for assessing scientific validity, but also noting that the Court did not provide specific guidance for evaluating these principles).
55. *Daubert*, 509 U.S. at 600–01 (Rehnquist, C.J., concurring in part and dissenting in part). Justice Blackmun and the majority did not share this concern. Id. at 592–93 (majority opinion) (“We are confident that federal judges possess the capacity to undertake this review.”).
56. See, e.g., Saks, supra note 30, at 144 (recognizing that some judges are unwilling or unable to apply *Daubert* in cases with complex scientific evidence); see also *Daubert* v. Merrell Dow Pharm., 43 F.3d 1311, 1315 (9th Cir. 1995) (noting that federal judges “face a far more complex and daunting task in a post-*Daubert* world than before”).
58. See *Kaye* et al., supra note 3, § 6.4.2, at 225 & n.14 (listing the courts that applied a *Daubert*-like standard before the decision).
59. See id. at 225 & n.14 (listing the courts that followed the Supreme Court and adopted the *Daubert* standard).
60. See supra note 54 and accompanying text (describing the judge’s role as gatekeeper under *Daubert* and the type of analysis performed by a judge in assessing the scientific validity of expert evidence).
admissibility standard for experts appears to have increased judicial scrutiny of scientific evidence.\textsuperscript{61}

\textbf{C. Prior Studies on Daubert: Surveys and Case Review Analysis}

Even if judges in both \textit{Daubert} and \textit{Frye} jurisdictions have examined scientific evidence more closely after \textit{Daubert}, many scholars question if the shift in standards had a major effect on the handling of cases. To help answer this question, researchers have both surveyed judges and performed quantitative analysis of reported cases.

\textit{1. Studies Relying on Survey Methodology}

Survey instruments since \textit{Daubert} have attempted to measure the effect of the Rule 702 standard on judicial practices. For example, a 2001 study by Sophia Gatowski and her coauthors analyzed state court judges’ perceptions of the gatekeeping function and the effect of \textit{Daubert}.\textsuperscript{62} Judges surveyed supported the gatekeeping role by a wide margin (91\%), and over 60\% of judges saw themselves taking an active role in admissibility analysis.\textsuperscript{63} When asked about the effect of \textit{Daubert} on the gatekeeping process, the judges’ responses did not show a clear consensus.\textsuperscript{64} The largest group of judges in the survey (36\%) believed \textit{Daubert} was not intended to raise or lower the threshold for admissibility, and an additional 11\% were unsure of \textit{Daubert’s} intent.\textsuperscript{65} Even among those who believed \textit{Daubert} did change the admissibility standard, the result was mixed. Twenty-three percent of respondents stated that the Court’s intent was to lower the threshold, while 32\% indicated that the court’s intent was to raise the standard.\textsuperscript{66} In that 2001 survey, then, judges lacked consensus on the effect of \textit{Daubert} on gatekeeping, even if many supported an active judicial role.

\begin{footnotesize}
\begin{enumerate}
\item See \textit{KAYE ET AL.}, supra note 3, at 186 (noting that \textit{Frye} jurisdictions apply the test with “enhanced vigor”); see also Bernstein, \textit{supra} note 35, at 393 (asserting that \textit{Frye} is “slowly converging with \textit{Daubert} jurisprudence”).
\item Gatowski et al., \textit{supra} note 2, at 433. The participating judges were from both \textit{Daubert} and \textit{Frye} jurisdictions. \textit{Id.} at 442 (surveying 400 judges: 205 from \textit{Daubert} jurisdictions and 195 from \textit{Frye} jurisdictions).
\item \textit{Id.} at 443 (noting that judges support the gatekeeping function regardless of which admissibility standard is applied).
\item Id. at 444 (finding a 14\% difference between judges who believed \textit{Daubert} changed their roles as gatekeeper and those who believed that their roles remained the same following the \textit{Daubert} decision).
\item Id. at 443 (stating that those who believed that \textit{Daubert} had no intent to change the standard for admissibility instead believed that \textit{Daubert’s} intent was to give judges the discretion to apply the admissibility framework themselves).
\item Id.
\end{enumerate}
\end{footnotesize}
In 2002, Carol Krafka and her colleagues surveyed federal district court judges to determine their opinions on the effect of *Daubert* in the courts.\(^{67}\) Krafka’s survey of over 300 respondents found that federal judges are more likely to closely scrutinize, limit, and exclude expert testimony after *Daubert*.\(^{68}\) Eighteen percent of the reported exclusions were due to the unreliability of the evidence.\(^{69}\) Krafka and her colleagues concluded that, since *Daubert*, judges more actively analyzed expert evidence and were less likely to admit it.\(^{70}\) However, the survey only included cases that went to trial.\(^{71}\)

In 2006, the National Center for State Courts (NCSC) published a preliminary study addressing the management of expert witnesses in Delaware state courts.\(^{72}\) In that study, NCSC researchers examined civil case filings and surveyed participants to assess the change in case management practices following the adoption of *Daubert* by Delaware courts in 1999.\(^{73}\) The study concluded that although the *Daubert* system in Delaware resulted in the judiciary seeking a more active gatekeeping role, ultimately “[t]he overall impact of *Daubert* has been minimal compared to the original fears when the U.S. Supreme Court issued its decisions on the admissibility of expert testimony.”\(^{74}\) Although judicial surveys published since *Daubert* uniformly support the proposition that judges now see themselves as more active in the management of complex scientific evidence, the studies are inconclusive on the critical question of the substantive effect of the *Daubert* standard. While Krafka’s study suggests that judges perceive themselves as more stringent under *Daubert*, Gatowski’s respondents were significantly more equivocal. Survey research has provided important insight on the issue of *Daubert*, but it leaves the central question unanswered.

\(^{67}\) Krafka et al., *supra* note 4, at 309 (noting that judges were asked about their most recent civil trial and their general experience with trials).

\(^{68}\) *Id.* at 322 (finding that judges assert more control over the admissibility of expert witness testimony post-*Daubert*).

\(^{69}\) *Id.* at 322–23 (explaining that, although exclusion of scientific evidence increased after *Daubert*, judges’ evaluation of the evidence largely depends on the reliability of the methodology of the expert rather than the other *Daubert* criteria, such as margin of error or that the method was subject to peer review).

\(^{70}\) *Id.* at 330 (concluding that the clarification of admissibility standards has motivated more careful consideration of proffered testimony).

\(^{71}\) *Id.* at 331 (“To determine how *Daubert* and its associated cases have affected judicial and attorney practices in the majority of cases that never go to trial, further research will be needed.”).

\(^{72}\) *Changing Role of Judges, supra* note 26, at 1–4.

\(^{73}\) *Id.* at 3 (examining civil case filings in Delaware Superior Court and interviewing 20 attorneys and 13 judges).

\(^{74}\) *Id.* at 3–4 (finding that *Daubert* did not result in an “excessive or unnecessary cost or delay” in Delaware courts).
2. Quantitative Analysis of Reported Cases

Researchers studying the effect of Daubert have also used non-survey methodologies, by analyzing reported court decisions to measure changes since 1993. Just as with the survey research, these studies are also largely inconclusive.

A 2001 study by Lloyd Dixon and Brian Gill examined the substantive effect of the Daubert decision in federal civil cases by analyzing challenges to expert evidence in federal district courts.\(^\text{75}\) Their methodology evaluated the number of reported decisions in Westlaw’s database that dealt with challenges to expert evidence and the results of reliability determinations in those decisions.\(^\text{76}\) Dixon and Gill concluded that the data “suggest that the standards for reliability tightened in the years after the Daubert decision.”\(^\text{77}\) However, the researchers noted that their study had two important limitations: (1) their findings did not address whether greater judicial scrutiny resulted in better outcomes in the affected cases; and (2) the study was not conclusive in determining whether Daubert’s effect was uniform across the judiciary.\(^\text{78}\) Even with the limitations, Dixon and Gill concluded that their “analysis of district court opinions suggests that following Daubert, judges scrutinized reliability more carefully and applied stricter standards in deciding whether to admit expert evidence.”\(^\text{79}\)

In 2002, Jennifer Groscup and her colleagues published an empirical study on the effect of Daubert on expert testimony in federal criminal cases.\(^\text{80}\) Using a methodology similar to Dixon and Gill’s analysis, Groscup used search terms in a computer database to see what trends or changes could be measured over time.\(^\text{81}\) In evaluating trends in both state and federal courts, Groscup found that “the basic rates of admission at the trial and the appellate court levels did not change significantly after Daubert in criminal cases on appeal.”\(^\text{82}\) Groscup suggested that the party offering the evidence, the type of counsel for the defense, and the standard of review may all play a part in the static rates of admission of expert evidence.\(^\text{83}\) Groscup concluded that, although there may be increased scrutiny of scientific evidence in criminal

\(^{75}\) DIXON & GILL, supra note 4, at xii–xiv.

\(^{76}\) Id. at 15–19 (explaining the methodology by which the data was collected).

\(^{77}\) Id. at 28 fig.4.1, 29 (reporting that the success rate for the challenge of expert evidence increased following Daubert).

\(^{78}\) Id. at 30–31 (noting that additional data is needed to address the study’s limitations).

\(^{79}\) Id. at 61.

\(^{80}\) GROSCUP et al., supra note 3, at 342 (examining appellate court decisions).

\(^{81}\) Id. at 342–44 (using specific terms and connectors and date restrictions in the Westlaw database).

\(^{82}\) Id. at 345 (noting that this conclusion was contrary to the prediction of most commentators).

\(^{83}\) Id. at 346–47.
cases, “no change in the overall rate of admission for all types of expert evidence was observed.”  

3. Prior Survey and Case Study Research—Conclusion

In assessing the results of prior research in the field, both survey instruments and case review studies remain largely inconclusive on the effect of the Daubert ruling. Clearly, the Krafka survey responses and the Dixon and Gill case review study demonstrate support for the conclusion that Daubert has raised the standard for admission of expert testimony. Conversely, Gatowski’s survey and the Groscup case review study demonstrate little to no difference between Frye and Daubert jurisdictions, suggesting the change is much ado about nothing.

Without a clear finding from prior research, and without more guidance from the Supreme Court, the question of Daubert’s effect remained open. To find out more, researchers Edward Cheng and Albert Yoon turned to a new tool of analysis.

II. REMOVAL RATE APPROACH TO ANALYZING DAUBERT

To answer the question of whether Daubert adopted a stricter standard for admitting scientific evidence, Cheng and Yoon adopted a new approach by analyzing a new metric and a large amount of case data to measure Daubert’s effect. Their study concluded that the adoption of the Daubert standard did not have a statistically significant effect.  

We found this conclusion counterintuitive and sought to re-examine the data Cheng and Yoon used, in order to try to explain their result. Surprisingly, our analysis, using the exact same datasets as Cheng and Yoon, returned the opposite result. We found that the adoption of Daubert did have a statistically significant effect when measured by the removal rate metric.

A. The Cheng and Yoon Study: Basic Method and Result

In their 2005 study, Cheng and Yoon’s innovation was to analyze state and federal aggregate civil case data, by isolating and then measuring the effect of the shift from the Frye standard to the Daubert standard by calculating the rate at which litigants remove cases with diversity jurisdiction from state court to federal court. After calculating this removal rate across several jurisdictions,
the study could then compare removal rates in *Frye* jurisdictions with removal rates in *Daubert* jurisdictions.\(^8^9\)

Because rate of removal is not reported by any given jurisdiction, Cheng and Yoon needed to calculate it themselves.\(^9^0\) To do so, they divided the number of tort cases removed from a jurisdiction’s state court and subsequently filed in that jurisdiction’s federal court by the number of tort cases in that state’s state court:\(^9^1\)

\[
\text{Rate of removal} = \frac{\text{# of tort cases removed to federal court}}{\text{# of tort cases filed in state court}}
\]

**Equation 1**

The numerator was calculated using data from the *Federal Court Cases: Integrated Data Base* (Federal Database) created by the Federal Judicial Center.\(^9^2\) The denominator was calculated using the *State Court Statistics, 1985–2001 Dataset*, which collected data from various state court databases (State Databases).\(^9^3\)

The central premise allowing for this comparison is that federal courts must use the Federal Rules of Evidence and, therefore, the *Daubert* standard is the only standard used in federal trials, including cases that originated in states using the *Frye* standard.\(^9^4\) As a result, a litigant in a *Frye* state can remove a case from the state court to the federal court, which is required to apply the *Daubert* standard, thus gaining the advantage of the different test for admissibility.\(^9^5\) Of course, litigants will only choose to remove their case if it

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89. *See Cheng & Yoon, supra note 6, at 482–84.*
90. *See id. at 488.* Unfortunately, Cheng and Yoon’s method of aggregating the data to calculate the removal rate has a number of flaws and produced data that could only be reproduced in an ad hoc fashion or could not be reproduced at all, which we discuss *infra* in Part II.D.
91. *See Cheng & Yoon, supra note 6, at 487 & fig.4* (describing the method by which Cheng and Yoon calculated the removal rate).
92. *Id. at 488* (describing the method by which Cheng and Yoon gathered their data).
93. *Id. at 488, 492.* Cheng and Yoon engaged in two related studies: the Preliminary Study and the National Study. The Preliminary Study utilized two state-specific databases: the New York data came from the Technology Division of the New York State Unified Court System database (New York Database), and the Connecticut data came from the Judicial information Systems division of the Connecticut Judicial Branch (Connecticut Database). *Id.* at 488. The National Study utilized a database created by the National Center for State Courts called *State Court Statistics, 1985 to 2001* (State Database). *Id.* at 492.
94. *See id. at 482–83* (discussing how admissibility standards are procedural and, therefore, do not have to be followed in federal court under the well-established *Erie Railroad Co. v. Tompkins* standard).
95. *See id.* (noting that the admissibility standard plays a major role in a litigant’s decision to remove a case to federal court).
is in their best interest. This innovative study design captured the effect of Daubert on a larger body of cases, not just those that went to trial or resulted in appellate decisions, as in previous studies. The new metric also avoided filtering the results through survey responses, allowing Cheng and Yoon to measure the real-world effect of the Daubert standard.

Cheng and Yoon compared the removal rate of 16 states—8 Daubert states and 8 Frye states—from 1994 to 2000. Using a fixed effects statistical model, the study found that “a state’s choice of scientific admissibility standard does not have a statistically significant effect on removal rates.” Because there is no effect of the standard on removal rates, Cheng and Yoon concluded that “a state’s adoption of Frye or Daubert makes no difference in practice” and that the power of Daubert lies solely in a judge’s scrutiny of expert evidence.

B. The Promise of Removal Rate Analysis

We believe that the removal rate metric is the most important innovation of the Cheng and Yoon study and that it offers significant benefits over alternative methods, such as survey work or case studies. The power of the removal rate approach lies in the scope of cases included in the analysis. Under the removal rate approach, we can capture Daubert’s effect on a larger body of cases, not just those that went to trial or resulted in appellate decisions. In our analysis infra, we could include an aggregate total of 3,997,970 cases. Case study research could never hope to individually

96. See id. at 508 (arguing that the study’s validity depends on defense counsel’s capacity to make decisions that are in their client’s best interest).
97. See id. at 484 (noting that the removal metric captures more cases because it considers an earlier stage of the litigation process, eliminating the role that confounding variables, such as selection bias and sealed settlements, play in the process).
98. Id. (arguing that the removal rate reduces the problem of inaccurate recall often encountered by survey research).
99. Id. at 492–94 (discussing the criteria for exclusion from the study, such as states that did not clearly adopt either Frye or Daubert or states with incomplete data).
100. For a detailed discussion of this procedure and the results, see infra Part II.D.2.a–d.
101. Cheng & Yoon, supra note 6, at 503.
102. Id. (emphasis added).
103. Id. at 503, 505 (suggesting that Daubert’s influence resulted in greater awareness of junk science).
104. Id. at 506 (“The removal rate metric offers an important, useful, and much-needed alternative.”); see also supra Part I.C (reviewing the inconclusive results achieved by alternative methods of evaluating Daubert’s effect).
105. See Dixon & Gilf., supra note 4, at xiii (analyzing only 399 district court opinions); see also Groscup et al., supra note 3, at 344 (analyzing only 693 appellate opinions).
106. This total refers to the cases collected from the State Database for the period from 1990 to 2000, which we analyzed in the National Study, infra Part II.D, the Flight-from-Frye analysis, infra Part II.C.2, and the Flight-to-State-Court analysis, infra Part II.C.3.
analyze such numbers, and we believe this larger sample improves the validity and reliability of our results.

Additionally, the removal rate approach avoids the potential shortcomings of quantitative survey research. Namely, the removal rate analysis measures the true test of litigants’ opinions—their actions when their self-interest is in play—rather than simply measuring survey responses.\(^{107}\)

Finally, as Cheng and Yoon noted in their work, the removal rate metric also removes one potential distorting factor from consideration.\(^{108}\) Because removal occurs early in the litigation process, the strength of the evidence in a specific case is likely unknown at the time of removal.\(^{109}\) As such, a litigant’s removal decision usually cannot be based on case-specific evidentiary concerns, but rather represents a general opinion of the relative merits of state or federal court approaches.\(^{110}\)

With these benefits, we are convinced that the removal rate metric offers the best opportunity to measure the true effect of the \textit{Daubert} decision.

\section*{C. Our Analysis of Removal Rates Shows That a State’s Adoption of \textit{Daubert} Does Effect the Rate of Removal}

Using a fixed-effects approach to removal-rate analysis, Cheng and Yoon concluded that a state’s adoption of \textit{Daubert} or \textit{Frye} does not affect removal rates in a statistically significant manner.\(^{111}\) Because this result was so counterintuitive, we set out to determine whether their conclusion could be confirmed. Our own statistical analysis demonstrates that defendants respond to a state’s adoption of \textit{Daubert} or \textit{Frye} in precisely the way one would intuitively expect if defendants think that the \textit{Daubert} standard is stricter than the \textit{Frye} standard.\(^{112}\)

\subsection*{1. Two Thought Experiments and the Expected Consequences of \textit{Daubert}’s Adoption: Flight-from-\textit{Frye} and Flight-to-State-Court}

To understand the true effect of \textit{Daubert} on removal rates, we must focus on what removal rates were intended to measure—defense attorneys’ views as to the relative strictness of the \textit{Daubert} and \textit{Frye} standards—and ask what the world would be like if defense attorneys believe that \textit{Daubert} is a stricter

\begin{itemize}
\item\(^{107}\) See Cheng & Yoon, \textit{supra} note 6, at 508 (assuming that attorneys remove cases to federal court when removal is in their client’s best interest).
\item\(^{108}\) \textit{Id.} at 484.
\item\(^{109}\) See 28 U.S.C. § 1446(b) (2006) (requiring that the removal decision be made within thirty days of receipt of the complaint).
\item\(^{110}\) Cheng & Yoon, \textit{supra} note 6, at 484 (arguing that litigants make the choice to remove to federal court when it is a more favorable forum, regardless of the relative strength of the underlying merits of the case).
\item\(^{111}\) \textit{Id.} at 503; see also \textit{supra} notes 101–03 and accompanying text.
\item\(^{112}\) See \textit{infra} Part II.C.1–4 (detailing our analysis, methodology, and conclusion).
\end{itemize}
standard. In engaging in this process we used two thought experiments. The first thought experiment, “Flight-from-Frye,” is based on two assumptions. First, if defense attorneys believe that Daubert is a stricter standard than Frye, then, in cases where the admissibility standard matters, they should remove the case from a Frye jurisdiction to a Daubert jurisdiction. Second, although there is a cost to remove a case from state court to federal court, it is not excessive.

If these assumptions are true, then we should expect to see the following: if state A is a Frye jurisdiction at time $t_1$ through $t_{10}$ and if the federal courts adopt Daubert at time $t_5$, then the rate of removals from state court to federal court should increase after time $t_5$, with a steep increase initially followed by gradual tapering. Graphically, this should look something like:

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113. See Cheng & Yoon, supra note 6, at 508 (arguing that removal rate analysis “depends on defense counsel’s judgment” of the “practical ramifications” of the admissibility standard in his or her jurisdiction); see also Henry G. Miller, The Daubert Debacle, N.Y. St. B.A. J., Mar./Apr. 2005, at 24, 28 (asserting that “[t]hose representing corporate defendants accused of manufacturing dangerous products will, of course, try to go to federal court and make as many Daubert motions as they can,” and that a defense attorney can subsequently win a case based on a Daubert motion alone); cf. David Paul Horowitz, “Will the Gatekeeper Let Daubert In?”, N.Y. St. B.J., June 2006, at 18, 18 (observing that defense attorneys in New York not only prefer to apply Daubert but have also advocated for its adoption in New York, a Frye state).

114. See 28 U.S.C. § 1914(a) (2006) (“The clerk of each district court shall require the parties instituting any civil action, suit or proceeding in such court, whether by original process, removal or otherwise, to pay a filing fee of $350.”).
Figure 1: The “Flight-from-Frye” Theoretical Effect in Years 1 to 10

This trend follows naturally from our first assumption: that defense attorneys believe that the Daubert standard will keep out more evidence than the Frye standard. Thus, when it matters, defense attorneys will remove a case to federal court more frequently after the federal courts adopt Daubert to take advantage of the stricter standard even if it costs something to do so. Moreover, the change will be an increase in the rate of removals.

Let us now reverse this in the “Flight-to-State-Court” thought experiment. Under the Flight-to-State-Court experiment, the two basic assumptions of Flight-from-Frye remain the same. Now, assume that state A is a Frye jurisdiction at time $t_1$ through $t_{10}$, the federal courts adopted Daubert at time $t_5$, but then state A adopts Daubert at time $t_{11}$. As noted in the Flight-from-Frye experiment, there should be an increase in the rate of removals occurring after time $t_5$. But, because state A adopts Daubert at time $t_{11}$, there is no longer any evidentiary advantage to removing from state court to federal court because the state and federal courts now will apply the same expert admissibility standard. Moreover, there is some disadvantage

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115. See supra notes 112–14 and accompanying text (articulating the Flight-from-Frye assumptions: (1) that defense attorneys will remove to federal court where they believe that the Daubert standard will be beneficial; and (2) that the cost of removal is manageable).

116. See supra FIGURE 1 and accompanying text (reporting the findings from the Flight-from-Frye experiment).
because, under our second assumption, there is a cost of removal. This implies that, given the rise in removal rates between time $t_5$ and $t_{10}$ caused by the federal courts’ adoption of Daubert when state A had not adopted Daubert, we should see a decrease in the rate of removals after time $t_{11}$ because of the lost evidentiary benefit of removal plus the cost of removal. Graphically, this should look something like:

![Figure 2: The “Flight-to-State-Court” Theoretical Effect in Years 1 to 15](image)

2. Analysis of Removal Rate Data Demonstrates Litigants Act in the Same Way as the Flight-from-Frye Thought Experiment

In the Flight-from-Frye experiment, we expect to see the rate of removals increase in states that do not change from Frye to Daubert during a period before and after the federal courts adopt Daubert. Our analysis clearly demonstrates this effect.

Our first step is to attempt to confirm (or deny) the Flight-from-Frye thought experiment. In doing so we are attempting to measure only the effect the federal courts’ adoption of Daubert has on removal rates. To measure this impact, we identified 13 states that were Frye states both before and for some

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117. 28 U.S.C. § 1914(a) (requiring a filing fee of $350 for removal to federal court).
118. See supra Figure 1 and accompanying text (reporting the findings from the Flight-from-Frye experiment).
period after the federal courts adopted *Daubert*. Using these states, we engaged in two identical analyses: one of pure federal data,\textsuperscript{120} and the other of mixed federal and state data.\textsuperscript{121}

In each analysis we began by reconstructing the base data. First, we calculated the rate of removal for each year in each state. Second, we used the rate of removal and the number of tort cases filed in the relevant court (federal or state) to create a dataset. The dataset contained one data entry for each case filed, and each entry contained a field for the year, state, and whether the case was or was not removed.\textsuperscript{122} This process created two datasets: a pure federal

\textsuperscript{119} We used the following *Frye* states: Alaska, see Contreras v. State, 718 P.2d 129, 134 (Alaska 1986) (using the *Frye* test and noting that the *Frye* test was adopted by *Pulakos v. State*, 476 P.2d 474, 478 (Alaska 1970), overruled by State v. Coon, 974 P.2d 386, 394 (Alaska 1999)); Arizona, see State v. Lehr, 38 P.3d 1172, 1178–79 (Ariz. 2002) (stating that “Arizona adheres to the *Frye* standard in ruling on the admissibility of novel scientific evidence”); Connecticut, see State v. Porter, 698 A.2d 739, 746 (Conn. 1997) (reconsidering Connecticut’s test for admitting scientific evidence, which had previously been based on the *Frye* test); Florida, see Williamson v. State, 994 So. 2d 1000, 1009 (Fla. 2008) (asserting that the *Frye* standard should be used before admitting expert testimony); Indiana, see Sears Roebuck & Co. v. Manuilo, 742 N.E.2d 453, 460–61 (Ind. 2001) (noting that Indiana courts previously relied on the *Frye* test); Kansas, see State v. Simmons, 254 P.3d 97, 101 (Kan. 2011) (asserting that opinions based on scientific methods or procedures must be scrutinized under the test articulated in *Frye*); Michigan, see Craig v. Oakwood Hosp., 684 N.W.2d 296, 305–06 (Mich. 2004) (noting that the testimony in question was subject to the *Frye* test); Minnesota, see State v. Nose, 649 N.W.2d 815, 818 (Minn. 2002) (using the *Frye* test); Missouri, see State v. Taylor, 298 S.W.3d 482, 500 (Mo. 2009) (stating that the *Frye* test was used to determine admissibility of scientific evidence); New York, see Giordano v. Market Am., Inc., 941 N.E.2d 727, 733 (N.Y. 2010) (stating that New York courts follow the *Frye* test); North Carolina, see State v. Peoples, 319 S.E.2d 177, 187 (N.C. 1984) (noting that *Frye* is the general rule for admissibility of expert testimony); Tennessee, see McDaniel v. CSX Transp., Inc., 955 S.W.2d 257, 262–65 (Tenn. 1997) (noting Tennessee’s prior adherence to *Frye*); and Washington, see State v. Roberts, 14 P.3d 713, 740–41 (Wash. 2000) (applying *Frye*).

\textsuperscript{120} See generally infra Part II.D (discussing the databases featured both in our study and in Cheng and Yoon’s). This approach is somewhat contrary to Cheng and Yoon’s, which focused solely on whether state adoption of *Daubert* makes a difference in state courts. See Cheng & Yoon, supra note 6, at 496–97 (stating that “our study only investigates whether *Frye* and *Daubert* makes a difference in state courts”). We believe examining pure federal data is advantageous because the total number of tort cases filed in federal courts is much smaller than in state courts; therefore, changes in the number of removed cases would have a larger effect on the removal rate.

\textsuperscript{121} We used the Federal Database to identify the number of tort cases removed from the identified states. See generally infra Part II.D.1.c (explaining the method by which we calculated removals to federal courts). Next, we used the State Database to identify the number of tort cases filed in each state. See generally infra Part II.D.1.d (discussing the method by which we calculated the number of tort cases filed in state courts).

\textsuperscript{122} This differs from Cheng and Yoon, who utilized rates of removal as their base data. See Cheng & Yoon, supra note 6, at 487–88. Thus, if we had a dataset with 5 years of data for 1 state, where each year had 100 entries, with a 5% removal rate each year, we would have a dataset with 500 entries of which 25 were identified as removed (5 per year). Cheng and Yoon would have 5 entries: 1 for each year, containing the rate of removal for that year.
dataset and a mixed federal and state dataset. The pure federal dataset contained entries using the Federal Database to calculate the number of cases removed, the number of tort cases filed, and the removal rate. The mixed dataset contained entries using the State Database to calculate the number of tort cases filed, the Federal Database to calculate the number of cases removed, and both of those numbers to calculate the removal rate.123

Starting with the pure federal dataset, we limited the data to include just those cases in the 13 states that were the appropriate kind of tort action from 1990 to 2000.124 Next, we limited the dataset to include only those tort cases that had been removed from state court to federal court.125 We excluded data from the specific states during certain periods of time where that data was excluded by Cheng and Yoon.126

It is important to note that we have assumed that state adoption of Daubert will result in a decrease in removal rates from the point in time when the state adopts Daubert—presuming the federal courts have already adopted Daubert.127 Therefore, to eliminate confusion that this counter-effect could have, we removed data associated with it by excluding data for each state starting from the year the state adopted Daubert.128 We also eliminated the year Daubert was adopted by the federal courts (1993) because the confusion of the transition year would likely disrupt the analysis.129

123. This second dataset is the same one used by Cheng and Yoon. See Cheng & Yoon, supra note 6, at 491–94 (noting that in the National Study, Cheng and Yoon used the same methodology for calculating removal rates as in the Preliminary Study).

124. See infra Part II.D.1.d.i (discussing in detail the method by which we chose the appropriate type of tort cases).

125. Because there are far fewer federal tort cases, the denominator (number of tort cases) is substantially smaller than the corresponding number of tort cases filed in state court.


127. See supra Figure 2 and accompanying text (explaining the Flight-to-State-Court experiment).

128. We excluded data for Alaska from 1999 onward, see State v. Coon, 974 P.2d 386, 394 (Alaska 1999) (holding that Alaska no longer applies the Frye test); Connecticut from 1997 onward see State v. Porter, 698 A.2d 739, 746 (Conn. 1997) (adopting Daubert); Indiana from 1995 onward, see Steward v. State, 652 N.E.2d 490, 498–99 (Ind. 1995) (assessing the admissibility of expert evidence based on the Indiana Rules of Evidence and the principles of Daubert); and Tennessee from 1997 onward, see McDaniel v. CSX Transp., Inc., 955 S.W.2d 257, 265 (Tenn. 1997) (adopting certain aspects of Daubert and allowing expert evidence only where it will “substantially assist the trier of fact” and does not “indicate a lack of trustworthiness”). Because Arizona, Florida, Kansas, Michigan, Minnesota, Missouri, New York, and Washington did not adopt Daubert in the relevant period, we used data from the entire relevant period for those states. See supra note 119 (detailing the standards for the states in question).

Using this data, we performed a fixed-effects analysis using logistic regression. This provided statistically significant results showing that the probability that a case in federal court was removed from state court increased after \textit{Daubert} was adopted.

<table>
<thead>
<tr>
<th>Daubert Adopted in U.S. Court</th>
<th>Probability of Removal</th>
</tr>
</thead>
<tbody>
<tr>
<td>No</td>
<td>0.259268</td>
</tr>
<tr>
<td>Yes</td>
<td>0.327401</td>
</tr>
</tbody>
</table>

\textbf{FIGURE 3: FIXED-EFFECTS ANALYSIS ON REMOVAL RATE USING FEDERAL DATA ONLY}

Next, we reproduced this analysis using the mixed Federal and State Database. This analysis also produced statistically significant results demonstrating that the adoption of \textit{Daubert} increases the probability of removal to federal court.

<table>
<thead>
<tr>
<th>Daubert Adopted in U.S. Court</th>
<th>Probability of Removal</th>
</tr>
</thead>
<tbody>
<tr>
<td>No</td>
<td>0.004517</td>
</tr>
<tr>
<td>Yes</td>
<td>0.005297</td>
</tr>
</tbody>
</table>

\textbf{FIGURE 4: FIXED-EFFECTS ANALYSIS ON REMOVAL RATE USING STATE AND FEDERAL DATA}

The difference between the probability of removal in Figure 3 and Figure 4—the probability of removal in Figure 3 is approximately 61.8 times larger

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130. See infra Part II.D.2.a–d (explaining in detail our fixed-effects model).

131. We used a logistic regression because this method is generally preferred when the independent variable is categorical/binary. See, e.g., ALAN AGRESTI & CHRISTINE FRANKLIN, STATISTICS 610 (2007); DAMODAR N. GUJARATI & DAWN C. PORTER, ESSENTIALS OF ECONOMETRICS 387–89 (2010).

132. All measures of statistical significance discussed in this Article relate to the \textit{p}-value of a statistical hypothesis. We will consider a result to be statistically significant if its corresponding \textit{p}-value is less than or equal to 0.05. This means that there is no more than a 1 in 20 chance (5% chance) that our result is due to chance. DAVID HENSHER, JOHN M. ROSE & WILLIAM H. GREENE, APPLIED CHOICE ANALYSIS: A PRIMER 46–47 (2005). In addition, a \textit{p}-value of \(\leq 0.05\) is consistent with general practice. See, e.g., \textit{id}; SCOTT E. MAXWELL & HAROLD D. DELANY, DESIGNING EXPERIMENTS AND ANALYZING DATA: A MODEL COMPARISON PERSPECTIVE 47 (2d ed. 2004).

133. Using only federal data, the \textit{p}-value was less than 0.0005. The results of the logistic regression are fairly robust and supported by the results of a linear regression, which provided a regression coefficient of 0.0595772, with a \textit{p}-value of less than 0.0005.

134. Using both federal and state data, the \textit{p}-value was less than 0.0005. Once again, the linear regression also supports the conclusion that the removal rate following the adoption of \textit{Daubert} in federal courts is statistically significant. Using linear regression in the fixed-effects analysis, we achieved a correlation coefficient of 0.0014486, with a \textit{p}-value of less than 0.0005.
than the probability of removal in Figure 4—is likely due to the difference in the method by which we calculated the rate of removal using the pure Federal Database and the mixed Federal and State Database. In both cases, we used the same removal rate formula (Equation 1), the same numerator (number of tort cases listed in the Federal Database as having been removed from state court), but different denominators (number of tort cases filed in state court versus number of tort cases filed in federal court in a given state). In the pure federal dataset, the denominator is the number of tort cases filed in federal court in a given state. In the mixed federal and state dataset, the denominator is the number of tort cases filed in a given state court. On average, there is only 1 tort case filed in federal court for every 30 tort cases filed in the relevant state court.135 Given this 30:1 ratio of filings, we should expect to find a comparable difference in probability of removal, which we did.136 These results unequivocally demonstrate that in states using the Frye standard, the adoption of Daubert by the federal courts results in an increased rate of removals to the federal courts.

3. Analysis of Removal Rate Data Also Demonstrates Litigants Act in the Same Way as the Flight-to-State-Court Thought Experiment

The Flight-to-State-Court thought experiment proposed that, if the state court was initially a Frye state, remained a Frye state for a period of time after the federal courts adopted Daubert, but eventually adopted Daubert, the rate of removals should decrease.137 Looking at actual case data, litigants’ actual behavior mirrored this result.138

We used 5 states that adopted Daubert from 1994 to 2000139—a period after the federal courts adopted Daubert—and 8 states that did not adopt Daubert

135. This ratio (30:1) was calculated using the 16 states used by Cheng & Yoon during the period from 1990 to 2000. See Cheng & Yoon, supra note 6, at 493 fig.7 (listing the states used by Cheng and Yoon during their National Study).

136. Compare supra FIGURE 3 and accompanying text (reporting the probability of removal using only federal data), with supra FIGURE 4 and accompanying text (reporting the probability of removal using both federal and state data).

137. See supra text accompanying notes 115–17.

138. See infra text accompanying notes 139–44; see also infra FIGURE 5.

139. The five states are Alaska, Connecticut, Indiana, North Carolina, and Tennessee. See State v. Coon, 974 P.2d 386, 394–95 (Alaska 1999) (adopting Daubert and rejecting Frye as inconsistent with the Alaska Rules of Evidence); see also State v. Porter, 698 A.2d 739, 746 (Conn. 1997) (holding that Daubert should be the new standard for accepting expert scientific evidence in the state); Steward v. State, 652 N.E.2d 490, 498 (Ind. 1995) (holding that, although the Daubert standard is not binding, it is helpful in applying the Indiana Rules of Evidence); State v. Goode, 461 S.E.2d 631, 639 (N.C. 1995) (holding that the test of whether to admit expert evidence requires analyzing the test adopted by the Supreme Court in Daubert); McDaniel v. CSX Transp., Inc., 955 S.W.2d 257, 262–65 (Tenn. 1997) (noting Tennessee’s prior adherence to Frye and holding that a partial acceptance of Daubert is appropriate). We ignore Arizona, Florida, Kansas, Michigan, Minnesota, Missouri, New York and Washington because they did not
through 2000 as controls.140 Once again, we engaged in two sequences of analysis: one using pure federal data and the other using mixed federal and state data.

Using the pure federal data, we performed a fixed-effects analysis using a logistic regression and produced a statistically significant result, at the $p \leq .05$ level, demonstrating that adoption of a Daubert standard by state courts after the federal courts results in a decreased probability of removal.141

<table>
<thead>
<tr>
<th>Daubert Adopted by State</th>
<th>Probability of Removal</th>
</tr>
</thead>
<tbody>
<tr>
<td>No</td>
<td>0.303158</td>
</tr>
<tr>
<td>Yes</td>
<td>0.25104</td>
</tr>
</tbody>
</table>

**FIGURE 5: FIXED-EFFECTS ANALYSIS ON REMOVAL RATE USING FEDERAL DATA ONLY**

Remarkably, the results in Figure 5 show that the rate of removal reverts to the levels of pre-federal adoption once the state also adopts Daubert.142 This is precisely what we expected to find under the Flight-to-State-Court thought experiment. Such a result strongly supports the underlying thesis that defense

adopt Daubert at any point during the duration of our study. See infra note 140. We also ignore Arkansas and Oregon because they adopted Daubert before the federal courts. See Prater v. State, 820 S.W.2d 429, 431 (Ark. 1991) (adopting a Daubert standard of admissibility); see also State v. Brown, 687 P.2d 751, 759 (Or. 1984) (adopting a liberal test that allows the judge to be a gatekeeper when determining whether certain evidence should be admitted). The tests adopted by both courts are similar to the test later adopted in Daubert. See Farm Bureau Mut. Ins. Co. v. Foote, 14 S.W.3d 512, 519 (Ark. 2000) (noting that Prater is “strikingly similar” to Daubert); see also Moore v. State, 915 S.W.2d 284, 294 (Ark. 1996) (comparing the approach adopted in Prater to the approach in Daubert); State v. O’Key, 899 P.2d 633, 680 (Or. 1995) (holding that Daubert is a further development of the process the court started in Brown). Finally, we ignore New Mexico because it adopted Daubert in 1993—the same time as the federal courts. See State v. Alberico, 861 P.2d 192, 203–04 (N.M. 1993) (rejected Frye and citing Daubert as the controlling standard of admissibility in New Mexico).


141. The $p$-value is 0.002. A fixed-effects analysis with linear regression did not provide a statistically significant result.

142. See supra Figure 3 (showing a similar rate of removal if the state and federal courts both adopt Daubert (.025104) as when the state and federal courts both did not adopt Daubert (0.259268)).
attorneys believe Daubert is the stricter standard.\textsuperscript{143} Next, we reproduced this analysis using the combined federal and state dataset but found no statistically significant relationships.\textsuperscript{144}

4. The Empirical Evidence of Actual Removal Rates Demonstrates Litigants Behavior After the Adoption of Daubert Mirrored Our Thought Experiments

By analyzing the removals from 1990 to 2000, we can demonstrate that those litigants in the dataset acted in the manner predicted by the Flight-from-Frye and Flight-to-State-Court thought experiments.\textsuperscript{145} This result strongly supports the conclusion that litigants themselves believe that the Daubert standard is stricter than the Frye standard.\textsuperscript{146}

D. Removal Rates Studies: Squaring the Circle

We began our project with the counterintuitive results from Cheng and Yoon’s work—that Daubert had no real effect on litigants and was much ado about nothing.\textsuperscript{147} Having run our analysis, the opposite seems to be true—litigants perceive Daubert to be a stricter standard.\textsuperscript{148} Both of these results, using the same data, cannot be correct. Therefore, it became necessary

\textsuperscript{143} See, e.g., Miller, supra note 113, at 24, 28 (explaining that defense attorneys remove cases to federal court when planning to file Daubert motions because Daubert employs a higher evidentiary standard than Frye).

\textsuperscript{144} Our suspicion is that the failure to find an effect is due to three factors. First, the number of torts filed in state court compared to the number of torts removed to federal courts is extremely small. See supra Figure 4 (noting that the rate of removal for torts filed in state courts is less than 0.53%, likely whether or not Daubert is adopted in federal courts). Second, the number of torts in which scientific evidence is involved is also very small. In this context, any change in the number of removals is likely to be very small compared to the total number of torts filed. Third, in this context, Indiana and North Carolina both adopted Daubert in 1995 and Alaska adopted Daubert in 1999. See State v. Coon, 974 P.2d 386, 394–95 (Alaska 1999); see also Steward v. State, 652 N.E.2d 490, 498 (Ind. 1995); State v. Goode, 461 S.E.2d 631, 639 (N.C. 1995). This means that for Indiana and North Carolina there is only one year where the trend can begin to be established before it is expected to reverse. Similarly, for Alaska there is only one year where the reversal would be measured. In such a context, it would almost be a miracle to find the effect.

\textsuperscript{145} See supra Part II.C.1–3.

\textsuperscript{146} After all, defense attorneys should remove their client’s case to federal court when it is in their client’s best interest. Cheng & Yoon, supra note 6, at 508. Therefore, the change in removal rates outlined in the previous sections supports the belief that Daubert is a stricter standard than Frye. See Miller, supra note 113, at 24, 28.

\textsuperscript{147} See Cheng & Yoon, supra note 6, at 503. For example, Cheng and Yoon suggest that, because the change in standard had no effect, state courts should consider uniformly adopting Daubert as the admissibility standard. Id. In addition, they suggest that advocates for reform should focus not on doctrinal tests, but on efforts to improve judicial outcomes. Id. at 504.

\textsuperscript{148} See supra note 112; supra Part II.C.1–4 and accompanying text (discussing how our study led to statistically significant results at the \( p < .05 \) level, demonstrating that whether a state court adopts Daubert significantly affects the removal rate to federal court).
so as to unpack the analysis Cheng and Yoon performed to find an explanation for the varying results. In so doing, we found three errors: (1) a miscalculation of removal rates; 149 (2) a failure to account for the difference in effect between state adoption of Daubert before federal adoption and state adoption of Daubert after federal adoption;150 and (3) a utilization of a study population that was too small to identify the effect of state adoption of Daubert.151 Combined, these three errors suggest that our results, rather than Cheng and Yoon’s, are correct; Daubert is the stricter standard.

1. Cheng & Yoon—Review of Data Aggregation

Before running a data analysis, a researcher must collect the underlying data to ensure its validity.152 Looking at the underlying data aggregation for the Cheng and Yoon study, we found what appear to be several errors.153 If their data is fundamentally flawed, any conclusions drawn from it will be undermined.

a. The Preliminary and National Studies

Cheng and Yoon’s removal rate analysis (the “Analysis”) occurred in two stages. In the first stage (the “Preliminary Study”), Cheng and Yoon compared Connecticut’s rate of removal in tort cases from state to federal court with removal rates154 from New York state court to the Eastern District of New York (EDNY) during the period 1994 to 2000.155 The Preliminary Study relied on a statistical model called “difference-in-differences,”156 a special case of fixed effect analysis.157 After engaging in a difference-in-differences analysis, Cheng and Yoon concluded that the 1997 adoption of Daubert by Connecticut did not have a statistically significant effect upon removal rates.158

149. See infra Part II.D.1.c.i.
150. See infra Part II.D.3.
151. See infra text accompanying notes 258–64, 269–70.
153. See supra text accompanying notes 149–51.
155. The EDNY includes the counties of Kings (the Bronx), Nassau, Queens, Richmond (Staten Island), and Suffolk. See EDNY, http://www.nycourts.gov/ (last visited Mar. 25, 2013) (stating “[t]he district comprises the counties of Kings, Nassau, Queens, Richmond, and Suffolk and concurrently with the Southern District, the waters within the counties of Bronx and New York”).
156. Cheng & Yoon, supra note 6, at 485–86 n.43.
157. See infra Part II.D.2.c (describing the difference-in-differences approach).
158. Cheng & Yoon, supra note 6, at 489–90. Cheng and Yoon fail to provide the result of the difference-in-differences analysis of the Preliminary Study and, instead, merely conclude that
In the second stage of the analysis (the “National Study”), Cheng and Yoon analyzed removal rates for 16 states\textsuperscript{159} from 1994 to 2000,\textsuperscript{160} utilizing a fixed effects statistical model.\textsuperscript{161} After multiple efforts at constructing a model, Cheng and Yoon concluded that “a state’s choice of scientific admissibility standard does not have a statistically significant effect on removal rates (or number of cases removed). This finding may support the broader theory that a state’s adoption of Frye or Daubert makes no difference in practice.”\textsuperscript{162}

\textit{b. The Target Data: Rate of Removal}

Because removal rate is not reported by any given jurisdiction, Cheng and Yoon calculated it themselves\textsuperscript{163} by dividing the number of tort cases removed from a jurisdiction’s state court and then filed in that jurisdiction’s federal court by the number of tort cases in that forum’s state court system:\textsuperscript{164}

\begin{equation}
\text{Rate of removal} = \frac{\# \text{ of tort cases removed to federal court}}{\# \text{ of tort cases filed in state court}}
\end{equation}

\textit{c. Calculating the Preliminary Study’s Rate of Removal}

The Preliminary Study attempted to assess Daubert’s impact on removal rates by comparing removal rates of the state of Connecticut with the EDNY from 1994 to 2000.\textsuperscript{165} In Cheng and Yoon’s study, the removal rate was calculated in six steps:

\textfootnote{159. These 16 states are Alaska, Arizona, Arkansas, Connecticut, Florida, Indiana, Kansas, Michigan, Minnesota, Missouri, New Mexico, New York, North Carolina, Oregon, Tennessee and Washington. Cheng & Yoon, supra note 6, at 492–93 (listing the states included in the National Study).}

\textfootnote{160. Id. at 492. Oddly, Cheng and Yoon label their study as “National Study, 1990–2000” even though the econometrics analysis is limited to the years 1994–2000. Id. They add the years 1990 through 1993 for “observational” purposes only, and do not include them in their analysis. Id.}

\textfootnote{161. Id. at 495–96 (stating that the model used the difference-in-differences approach, a model that assumes the observed variables are non-random).}

\textfootnote{162. Id. at 496–503.}

\textfootnote{163. See, e.g., id. at 487.}

\textfootnote{164. See id.}

\textfootnote{165. Id. at 485. Cheng and Yoon believed that such a comparison was appropriate for a number of reasons. First, New York and Connecticut are geographically close, economically connected to New York City, and “demographically similar.” See id. at 486. This implies that removal rate (and any subsequent increase or decrease in removal rates) of both areas, all else being equal, should be similar. Second, at the time, New York state courts were Frye jurisdictions while Connecticut state courts transitioned from Frye to Daubert in 1997. See State
1. Calculate the number of torts in Connecticut state court using the Connecticut database;\textsuperscript{166}
2. Calculate the number of torts in New York state court for the geographic area coextensive with the EDNY using the New York database;\textsuperscript{167}
3. Calculate the number of removals from the Connecticut state courts to federal court that were filed in the United States District Court for the District of Connecticut;\textsuperscript{168}
4. Calculate the number of removals from the relevant New York state courts to the United States District Court for the EDNY;\textsuperscript{169}
5. Use Equation 1 to calculate removal rate for New York;\textsuperscript{170} and
6. Use Equation 1 to calculate removal rate for Connecticut.\textsuperscript{171}

According to Cheng and Yoon, these steps produce the following results:\textsuperscript{172}

\vspace{1cm}
\textsuperscript{166} Cheng & Yoon, supra note 6, at 485–86.
\textsuperscript{167} See id.
\textsuperscript{168} See id.
\textsuperscript{169} See id.
\textsuperscript{170} See id. at 487–88.
\textsuperscript{171} See id.
\textsuperscript{172} Id. at 489.
TABLE 1

<table>
<thead>
<tr>
<th>Year</th>
<th># of torts filed</th>
<th># of cases removed to federal court</th>
<th>% Rate-of-Removal</th>
<th># of torts filed</th>
<th># of cases removed to federal court</th>
<th>% Rate-of-Removal</th>
</tr>
</thead>
<tbody>
<tr>
<td>1994</td>
<td>16,172</td>
<td>56</td>
<td>0.35</td>
<td>42,120</td>
<td>207</td>
<td>0.49</td>
</tr>
<tr>
<td>1995</td>
<td>18,417</td>
<td>64</td>
<td>0.35</td>
<td>46,199</td>
<td>237</td>
<td>0.51</td>
</tr>
<tr>
<td>1996</td>
<td>20,165</td>
<td>48</td>
<td>0.24</td>
<td>47,711</td>
<td>333</td>
<td>0.70</td>
</tr>
<tr>
<td>1997</td>
<td>20,295</td>
<td>49</td>
<td>0.24</td>
<td>47,235</td>
<td>263</td>
<td>0.70</td>
</tr>
<tr>
<td>1998</td>
<td>20,054</td>
<td>63</td>
<td>0.31</td>
<td>46,808</td>
<td>288</td>
<td>0.62</td>
</tr>
<tr>
<td>1999</td>
<td>18,845</td>
<td>52</td>
<td>0.28</td>
<td>45,838</td>
<td>310</td>
<td>0.68</td>
</tr>
<tr>
<td>2000</td>
<td>18,201</td>
<td>56</td>
<td>0.31</td>
<td>43,964</td>
<td>362</td>
<td>0.82</td>
</tr>
</tbody>
</table>

**d. Recreating Rate of Removal for the Preliminary Study**

**i. The First Problem: Calculating the Number of Removals to Federal Court**

The first step in our review of Cheng and Yoon’s work was to attempt to confirm the results of the Preliminary Study. First, we recreated Table 1 using the same data as Cheng and Yoon.174 The Federal Database came with a set of codebooks that explained what fields were included in the data and what each meant.175 The first field of interest in the Federal Database is the “nature of

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173. Because the number of removed cases is generally much smaller than the number of filed cases, we will report rate of removal as a percentage removal rate. Thus, a removal rate of 0.35% corresponds to a removal rate of 0.0035.

174. See infra Part II.D.1.d.iii. Cheng and Yoon graciously provided us with a copy of the data used in the study. When we were unable to recreate Cheng and Yoon’s aggregation numbers from that data, the authors obtained a copy of the Federal Database directly from the Federal Judicial Center. That database matched the one Cheng and Yoon provided.

175. Codebooks were provided for various periods of time. See Fed. Judicial Ctr., Federal Court Cases: Integrated Data Base, 1970–2000 pt. 117 [hereinafter Fed. Judicial Ctr. Data Base], available at http://www.icpsr.umich.edu/icpsrweb/NACJD/studies/08429/ascii (covering 2000); see also id. at pt. 116 (covering 1999); id. at pt. 115 (covering 1998); id. at pt. 104 (covering 1997); id. at pt. 103 (covering 1996); id. at pt. 98 (covering 1995); id. at pt. 88 (covering 1994); id. at pt. 87 (covering 1993); id. at pt. 86 (covering 1992); id. at pt. 74 (covering 1991); id. at pt. 73 (covering 1990); id. at pt. 65 (covering 1989); id. at pt. 64 (covering 1988). All codebooks are on file with the authors.
suit” field. According to the codebooks, a case was a tort if it had a nature of suit value of one of the following:

240 TORTS TO LAND
245 TORT PRODUCT LIABILITY
310 AIRPLANE PERSONAL INJURY
315 AIRPLANE PRODUCT LIABILITY
320 ASSAULT, LIBEL, AND SLANDER
330 FEDERAL EMPLOYERS’ LIABILITY
340 MARINE PERSONAL INJURY
345 MARINE - PRODUCT LIABILITY
350 MOTOR VEHICLE PERSONAL INJURY
355 MOTOR VEHICLE PRODUCT LIABILITY
360 OTHER PERSONAL INJURY
362 MEDICAL MALPRACTICE
365 PERSONAL INJURY - PRODUCT LIABILITY
368 ASBESTOS PERSONAL INJURY - PROD.LIAB.
370 OTHER FRAUD
371 TRUTH IN LENDING
380 OTHER PERSONAL PROPERTY DAMAGE
385 PROPERTY DAMAGE - PRODUCT LIABILITY

The second field of interest is the “district” field, which identifies the district court where the case was filed. The third field of interest is the “origin” field, which would be a “2” if the case had been removed from state court to federal district court.

These fields were used to restrict the dataset in the Federal Database to torts, removed from state court to federal court, for the District Court of Connecticut and for the EDNY. Next, we calculated the number of torts removed to federal

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176. See id. at pt. 117, at 7.
177. See id. at pt. 117, at 8.
178. See, e.g., id. at pt. 117, at 5.
179. See, e.g., id. The district field for District of Connecticut was “05” and the district field for New York cases filed in the EDNY was “07.” Id.
180. See, e.g., id. at pt. 117, at 6.
181. Id. at pt. 117, at 6–7. The origin field indicates how the case was filed in the district court and each code corresponds to a particular path the case may have followed to reach the district court. See id.
court in the EDNY and District Court for Connecticut, producing results that
differed significantly from those provided by Cheng and Yoon:182

<table>
<thead>
<tr>
<th>Year</th>
<th>Cheng &amp; Yoon removals</th>
<th>Authors’ # of removals</th>
<th>Difference</th>
<th>Cheng &amp; Yoon removals</th>
<th>Authors’ # of removals</th>
<th>Difference</th>
</tr>
</thead>
<tbody>
<tr>
<td>1994</td>
<td>56</td>
<td>73</td>
<td>-17</td>
<td>207</td>
<td>259</td>
<td>-52</td>
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<td>64</td>
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<td>358</td>
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<td>1996</td>
<td>48</td>
<td>71</td>
<td>-23</td>
<td>333</td>
<td>414</td>
<td>-81</td>
</tr>
<tr>
<td>1997</td>
<td>49</td>
<td>68</td>
<td>-19</td>
<td>263</td>
<td>319</td>
<td>-56</td>
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<td>1998</td>
<td>63</td>
<td>87</td>
<td>-24</td>
<td>288</td>
<td>340</td>
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<td>56</td>
<td>73</td>
<td>-17</td>
<td>362</td>
<td>413</td>
<td>-51</td>
</tr>
</tbody>
</table>

**TABLE 2**

Because of the significant differences in results, we reviewed the data to
identify where Cheng and Yoon may have reasonably further limited the data
to arrive at their results. Eventually, we determined it was the jurisdiction
field, which states:183

1 -US government plaintiff
2 -US government defendant
3 -federal question
4 -diversity of citizenship
5 -local question

We determined that Cheng and Yoon must have limited jurisdiction to only
diversity of citizenship184 because doing so eliminates many of the differences
between our results and their results:

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182. See Cheng & Yoon, supra note 6, at 489 fig.5.
183. See, e.g., FED. JUDICIAL CTR. DATA BASE, supra note 175, at pt. 117, at 7.
184. Id. Cheng and Yoon likely ignored the “1,” “2,” and “3” field values because such cases
were likely to be removed for reasons other than scientific evidence admissibility standards. We
also assume the “local question” cases, field value of “5,” were automatically ignored because
there were no cases that fit into category “5” in the relevant period.
### TABLE 3

Although the differences in results were smaller after adjusting for the jurisdiction limitation, they still did not match.\(^ {\text{185}}\) Through trial and error, we were able to (mostly) conform our results to Cheng and Yoon’s results by eliminating three categories of torts: products liability, torts to land, and assault, libel, and slander.\(^ {\text{186}}\)

<table>
<thead>
<tr>
<th>Year</th>
<th>Connecticut C&amp;Y # of removals</th>
<th>Authors’ # of removals</th>
<th>Difference</th>
<th>New York (limited to EDNY) C&amp;Y # of removals</th>
<th>Authors’ # of removals</th>
<th>Difference</th>
</tr>
</thead>
<tbody>
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<td>-4</td>
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<td>212</td>
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<td>67</td>
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<td>-3</td>
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<td>49</td>
<td>-1</td>
<td>333</td>
<td>338</td>
<td>-5</td>
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<td>263</td>
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<tr>
<td>1998</td>
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<td>68</td>
<td>-5</td>
<td>288</td>
<td>289</td>
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</tr>
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<td>-4</td>
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<tr>
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<td>56</td>
<td>58</td>
<td>-2</td>
<td>362</td>
<td>363</td>
<td>-1</td>
</tr>
</tbody>
</table>

### TABLE 4

Removing these three tort categories was an ad hoc measure, but it did produce results that conformed to almost all of Cheng and Yoon’s reported results.\(^ {\text{187}}\) The only remaining difference was New York, in 2000, where we

---

185. See supra TABLE 3 (showing a disparity in the results even after the jurisdictional limitation was adjusted).

186. The “nature of suit” field was restricted to values between 300 and 400, excluding cases with the following values: 240 (torts to land), 245 (tort product liability), and 320 (assault, libel, and slander). See, e.g., FED. JUDICIAL CTR. DATA BASE, supra note 175, pt. 117, at 7–8.

187. See supra TABLE 4 (showing more comparable results than in TABLE 3).
found 5 fewer torts.\textsuperscript{188} That result would match but only if we reintroduced assault, libel, and slander torts \textit{for just New York in 2000}.\textsuperscript{189}

Our attempt to recreate Cheng and Yoon’s removals count to federal court identified a substantial methodological flaw in their calculations. We could only reproduce their results by adopting two different ad hoc approaches.\textsuperscript{190} Moreover, removing product liability torts from the analysis is of grave concern in a study on the effect of a change in scientific evidence admissibility standards because product liability cases are the torts where scientific evidence is most likely to be present and contested.\textsuperscript{191} Further, there is no rational basis for removing assault, libel, and slander, or torts to land (or reintroducing assault, libel, and slander for just New York in 2000).\textsuperscript{192}

\textbf{ii. Calculating the Number of Torts in State Court}

Although Cheng and Yoon’s data aggregation from the Federal Database did not match our aggregation without ad hoc data manipulation, we were able to replicate Cheng and Yoon’s data aggregation of the Connecticut database and New York database to calculate the number of torts filed in each state.\textsuperscript{193}

\begin{itemize}
  \item \textsuperscript{188} See supra TABLE 4.
  \item \textsuperscript{189} However, if we reintroduced assault, libel, and slander to match Cheng and Yoon’s results, the only data points that would match would be New York in 2000.
  \item \textsuperscript{190} See supra notes 184, 186 and accompanying text (explaining that we limited the data by jurisdiction and removed three “nature of suit” categories to match over 92% of Cheng and Yoon’s results). Cheng and Yoon do not provide any explanation for their data aggregation methods that would account for or explain these differences. Rather they simply stated that “we extracted only tort cases filed in the EDNY and the District of Connecticut for the period of 1994 to 2000” and that they report this data in raw form in their Figure 5. Cheng & Yoon, supra note 6, at 488–89 & fig.5 (“Figure 5 shows the raw numbers and the calculated removal rates for Connecticut and the EDNY.” (emphasis added)).
  \item \textsuperscript{191} See Cheng & Yoon, supra note 6, at 483–84 (discussing the importance of scientific evidence in products liability tort cases); see also Margaret A. Berger, Upsetting the Balance Between Adverse Interests: The Impact of the Supreme Court’s Trilogy on Expert Testimony in Toxic Tort Litigation, 64 LAW & CONTEMP. PROBS. 289, 289–90 (2001) (explaining the contested nature of scientific evidence in toxic tort cases and how those cases led the fight for a reformed standard).
  \item \textsuperscript{192} See Cheng & Yoon, supra note 6, at 485–91 (failing to provide any explanation for their methods for aggregating data).
  \item \textsuperscript{193} There were two minor differences. First, in 1997, we counted 20,294 torts filed in Connecticut while Cheng and Yoon counted 20,295 torts filed in Connecticut. See id. at 489 fig.5; see also supra TABLE 1. Second, in 1999, we counted 18,844 torts filed in Connecticut while Cheng and Yoon counted 18,845 torts filed in Connecticut. See Cheng & Yoon, supra note 6, at 489 fig.5; see also supra TABLE 1. We presume that this is a reporting or typographical error.
\end{itemize}
iii. Because of Errors in Their Process, It Is Not Possible to Recreate Cheng and Yoon’s Preliminary Study

Although we were able to essentially recreate Cheng and Yoon’s calculation of the number of torts filed in Connecticut and New York, we could not match Cheng and Yoon’s calculation of the number of removals to federal district court. Consequently, we reject Cheng and Yoon’s calculation of the removal rate from 1994 to 2000 in Connecticut and the EDNY.

e. Recreating the Data for the National Study

Cheng and Yoon’s National Study examined the removal rate in 16 states from 1994 to 2000. Their calculation of those rates of removal for that period are provided below:

---

194. See supra Part II.D.1.d.i (discussing the difficulty in determining how Cheng and Yoon calculated the number of removals from state to federal court during the Preliminary Study).


196. Id. at 491–92 (noting the econometrics analysis was from 1994 to 2000).

197. Id. at app. a at 512–13; see also infra TABLE 5.
TABLE 5

As with the Preliminary Study, our first step was to attempt to recreate Cheng and Yoon’s removal rates. Here, recreating the removal rates was more problematic because, unlike in the Preliminary Study, we could not find any method that would reproduce all of Cheng and Yoon’s removal rates as reported in the National Study.198

To reproduce Cheng and Yoon’s National Study, we used the Federal Database to calculate the number of torts removed from state court that were filed in the appropriate federal court. Initially, we limited the set of cases to torts, from 1990 to 2000, that were removed to federal court under diversity jurisdiction. Then, we calculated the number of torts in each state using the State Database. This calculation produced the following removal rates:

<table>
<thead>
<tr>
<th>State</th>
<th>90</th>
<th>91</th>
<th>92</th>
<th>93</th>
<th>94</th>
<th>95</th>
<th>96</th>
<th>97</th>
<th>98</th>
<th>99</th>
<th>00</th>
</tr>
</thead>
<tbody>
<tr>
<td>Alaska</td>
<td>1.69</td>
<td>2.39</td>
<td>5.15</td>
<td>1.39</td>
<td>1.49</td>
<td>1.17</td>
<td>1.79</td>
<td>2.39</td>
<td>2.83</td>
<td>1.62</td>
<td>1.94</td>
</tr>
<tr>
<td>Arizona</td>
<td>0.3</td>
<td>0.36</td>
<td>0.51</td>
<td>0.61</td>
<td>0.26</td>
<td>0.25</td>
<td>0.31</td>
<td>0.4</td>
<td>0.34</td>
<td>0.31</td>
<td>0.48</td>
</tr>
<tr>
<td>Arkansas</td>
<td>1.27</td>
<td>1.12</td>
<td>1.35</td>
<td>1.42</td>
<td>1.28</td>
<td>0.57</td>
<td>1.45</td>
<td>1.35</td>
<td>1.87</td>
<td>2.09</td>
<td>1.75</td>
</tr>
<tr>
<td>Connecticut</td>
<td>0.16</td>
<td>0.23</td>
<td>0.29</td>
<td>0.24</td>
<td>0.24</td>
<td>0.36</td>
<td>0.36</td>
<td>0.25</td>
<td>0.25</td>
<td>0.31</td>
<td>0.28</td>
</tr>
<tr>
<td>Florida</td>
<td>0.58</td>
<td>0.73</td>
<td>0.61</td>
<td>0.54</td>
<td>0.43</td>
<td>0.31</td>
<td>0.46</td>
<td>0.76</td>
<td>0.58</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Indiana</td>
<td>0.86</td>
<td>0.86</td>
<td>1.45</td>
<td>1.13</td>
<td>0.51</td>
<td>0.55</td>
<td>0.91</td>
<td>1.17</td>
<td>0.85</td>
<td>1.13</td>
<td>1.01</td>
</tr>
<tr>
<td>Kansas</td>
<td>0.52</td>
<td>0.76</td>
<td>1.34</td>
<td>0.89</td>
<td>0.65</td>
<td>0.43</td>
<td>0.66</td>
<td>0.79</td>
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<td>0.76</td>
<td>0.86</td>
</tr>
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<td>0.64</td>
<td>0.49</td>
<td>0.49</td>
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<td>0.97</td>
<td>0.98</td>
<td>0.96</td>
<td>0.98</td>
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<tr>
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<td>1.97</td>
<td>0.98</td>
<td>0.9</td>
<td>0.86</td>
<td>0.48</td>
<td>0.8</td>
<td>1.87</td>
<td>0.99</td>
<td>0.92</td>
<td>1.12</td>
</tr>
<tr>
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<td>1.08</td>
<td>1.1</td>
<td>0.66</td>
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<td>0.81</td>
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<td>0.99</td>
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<td></td>
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<td>0.8</td>
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</tr>
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<td>0.56</td>
<td>0.42</td>
<td>0.37</td>
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<td>0.62</td>
<td>0.64</td>
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<td>0.97</td>
<td>1.08</td>
<td>1.46</td>
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<td>0.48</td>
<td>0.86</td>
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</tr>
</tbody>
</table>

198. See supra note 187 and accompanying text (discussing the ad hoc method we used to reproduce most of Cheng and Yoon’s removal rates in the Preliminary Study).
<table>
<thead>
<tr>
<th>State</th>
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<td>1.51</td>
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<td>1.14</td>
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<td>1.25</td>
<td>0.88</td>
<td>1.13</td>
<td>1.03</td>
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<tr>
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<td>0.93</td>
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<td>1.02</td>
<td>0.87</td>
<td>0.84</td>
<td>0.80</td>
<td>0.80</td>
<td>0.99</td>
</tr>
<tr>
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<td>0.67</td>
<td>0.65</td>
<td>1.01</td>
<td>0.97</td>
<td>1.02</td>
<td>1.00</td>
<td>0.99</td>
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<td>1.08</td>
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<td>1.01</td>
<td>0.97</td>
<td>1.19</td>
</tr>
<tr>
<td>Missouri</td>
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<td>1.18</td>
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<td>1.26</td>
<td>1.06</td>
<td>1.30</td>
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<td>1.17</td>
<td>1.14</td>
</tr>
<tr>
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<td>1.65</td>
<td>1.25</td>
<td>1.08</td>
<td>1.01</td>
<td>0.99</td>
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<td></td>
<td></td>
</tr>
<tr>
<td>New York</td>
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<td>0.59</td>
<td>0.69</td>
<td>0.86</td>
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<td>0.80</td>
<td>0.80</td>
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<tr>
<td>North Carolina</td>
<td>0.35</td>
<td>0.67</td>
<td>0.51</td>
<td>0.63</td>
<td>0.60</td>
<td>0.75</td>
<td>0.72</td>
<td>0.56</td>
<td>0.68</td>
<td>0.67</td>
<td>0.80</td>
</tr>
<tr>
<td>Oregon</td>
<td>0.67</td>
<td>0.73</td>
<td>0.87</td>
<td>1.22</td>
<td>1.06</td>
<td>1.14</td>
<td>0.72</td>
<td>0.42</td>
<td>0.60</td>
<td>0.71</td>
<td></td>
</tr>
<tr>
<td>Tennessee</td>
<td>1.09</td>
<td>0.87</td>
<td>1.19</td>
<td>1.34</td>
<td>1.67</td>
<td>1.67</td>
<td>1.37</td>
<td>1.46</td>
<td>1.32</td>
<td>1.40</td>
<td>1.72</td>
</tr>
<tr>
<td>Washington</td>
<td>0.65</td>
<td>0.25</td>
<td>0.43</td>
<td>0.44</td>
<td>0.35</td>
<td>0.32</td>
<td>0.46</td>
<td>0.24</td>
<td>0.27</td>
<td>0.33</td>
<td>0.50</td>
</tr>
</tbody>
</table>

**TABLE 6**
Once again, our results differed from Cheng and Yoon’s:\footnote{See \textit{infra} Table 7. \textit{Table 7} was compiled by subtracting the values Cheng and Yoon found from the values we found.} We were only able to find 15 data points (state-by-year) out of the 170 total\footnote{See \textit{supra} Table 7. \textit{Table 7} has a total of 176 entries (16 states by 11 years). Of these, we leave 6 blank (Florida in 1999 and 2000; New Mexico in 1990, 1991, and 1992; Oregon in 1990) because Cheng and Yoon also left those dates blank. Cheng & Yoon, \textit{supra} note 6, app. a, at 512–13; see also \textit{supra} Table 7. Cheng and Yoon did not explain why those entries were left blank. Cheng & Yoon, \textit{supra} note 6, at 491–513 (failing to account for blank entries). Although they did indicate that the databases used to calculate these results may have been either inaccurate or incomplete. \textit{Id.} at 492.} where our results align with Cheng and Yoon’s results—a 9% success rate. Then, we attempted to locate a method for counting that enabled us to more closely reproduce the results of Cheng and Yoon’s calculation of removal.

\begin{table}[h]
\centering
\begin{tabular}{|l|cccccccccc|}
\hline
& 90 & 91 & 92 & 93 & 94 & 95 & 96 & 97 & 98 & 99 & 00 \\
\hline
Alaska & -0.32 & -0.47 & -1.50 & -0.29 & 0.36 & 0.24 & -0.28 & 0.00 & 0.19 & 0.00 & 0.11 \\
Arizona & 0.02 & 0.02 & 0.04 & 0.04 & 0.08 & 0.16 & 0.05 & 0.02 & 0.03 & 0.00 & 0.00 \\
Arkansas & 0.06 & 0.12 & 0.04 & 0.09 & 0.49 & 0.84 & 0.21 & 0.09 & 0.09 & 0.10 & 0.14 \\
Connecticut & 0.02 & 0.00 & 0.04 & 0.00 & 0.02 & 0.01 & 0.01 & 0.02 & 0.03 & 0.02 & 0.01 \\
Florida & -0.31 & -0.45 & -0.38 & -0.34 & -0.24 & -0.11 & -0.24 & -0.46 & -0.35 & 0.00 & 0.00 \\
Indiana & -0.25 & -0.01 & -0.01 & 0.00 & 0.16 & 0.59 & 0.16 & 0.08 & 0.03 & 0.00 & 0.02 \\
Kansas & 0.05 & 0.07 & 0.04 & 0.04 & 0.14 & 0.59 & 0.21 & 0.05 & 0.01 & 0.04 & 0.13 \\
Michigan & 0.07 & 0.04 & 0.01 & 0.03 & 0.16 & 0.52 & 0.26 & 0.05 & 0.02 & 0.03 & 0.04 \\
Minnesota & 0.23 & 0.10 & 0.08 & 0.08 & 0.25 & 0.60 & 0.20 & 0.04 & 0.02 & 0.05 & 0.07 \\
Missouri & 0.15 & 0.04 & 0.04 & 0.10 & 0.30 & 0.60 & 0.28 & 0.06 & 0.04 & 0.03 & 0.03 \\
New Mexico & & & 0.00 & 0.10 & 1.28 & 0.44 & 0.05 & 0.00 & 0.00 & 0.05 & 0.00 \\
New York & 0.02 & 0.03 & 0.04 & 0.08 & 0.15 & 0.23 & 0.07 & 0.00 & 0.00 & 0.01 & 0.00 \\
North Carolina & 0.04 & 0.18 & 0.04 & 0.07 & 0.18 & 0.38 & 0.28 & 0.04 & 0.06 & 0.03 & 0.03 \\
Oregon & -0.21 & -0.24 & -0.21 & -0.24 & 0.45 & 0.66 & -0.14 & 0.01 & 0.04 & -0.03 & 0.00 \\
Tennessee & 0.12 & 0.03 & 0.04 & 0.10 & 0.42 & 0.91 & 0.34 & 0.07 & 0.04 & 0.04 & 0.05 \\
Washington & 0.04 & 0.01 & 0.00 & 0.06 & 0.07 & 0.16 & 0.08 & 0.01 & 0.00 & 0.02 & 0.01 \\
\hline
\end{tabular}
\caption{Table 7}
\end{table}
rate. Using the same ad hoc measures used in the Preliminary Study, we were able to produce results that were closer to Cheng and Yoon’s:

<table>
<thead>
<tr>
<th></th>
<th>90</th>
<th>91</th>
<th>92</th>
<th>93</th>
<th>94</th>
<th>95</th>
<th>96</th>
<th>97</th>
<th>98</th>
<th>99</th>
<th>00</th>
</tr>
</thead>
<tbody>
<tr>
<td>Alaska</td>
<td>-0.40</td>
<td>-0.70</td>
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Table 8

Although this analysis does allow us to align 66 data points with Cheng and Yoon, that only comprises 39% of the entries. Ultimately, we were not able to find a method of calculating removal rates that produced results that matched Cheng and Yoon’s table of removal rates. Thus, we can only conclude that a set of errors crept into their data calculation techniques.

f. Summary and Data Aggregation Conclusions

In order to perform a useful statistical analysis, one must input data that is consistent and understandable. Unfortunately, the data aggregation that Cheng and Yoon used to create the inputs to their statistical analysis is fundamentally flawed. Thus, any conclusions drawn from their data is suspect.

201. See supra notes 184–86 and accompanying text.
202. See supra Table 8.
2. Recreating the Preliminary Study: A Statistically Significant Result Found Once the Proper Study Population Is Used

As previously noted, Cheng and Yoon’s calculation of removal rates either miscalculates the rate of removal or utilizes an unknown, unidentified, and non-systematic method of calculation.\(^{203}\) Recreating the Preliminary Study using our own methods and our rates of removal allows us to conclude that there is a statistically significant effect from Connecticut’s adoption of Daubert in 1997.\(^{204}\) Cheng and Yoon’s failure to find a statistically significant result was likely due to the use of removal rates as the study population. We find statistical significance by using properly calculated removal rates, counts of torts filed, and counts of torts removed to create a dataset in which there is one entry for each tort filed in the respective court and each data entry is identified as either removed or not based on the removal rate.\(^{205}\)

a. An Intuitive Explanation of the Difference-in-Differences Analysis

Cheng & Yoon attempt to analyze whether the rate of removal is correlated in a meaningful way with state adoption of Daubert by utilizing a difference-in-differences approach.\(^{206}\) A difference-in-differences analysis can be seen as combining two comparison techniques: before-after analysis and matching.\(^{207}\) In a before-after analysis, we measure the effect a treatment\(^{208}\) has on a population by comparing the tested variable\(^{209}\) before and after the treatment is given,\(^{210}\) for example, comparing the removal rate in Connecticut before it adopts Daubert (pre-1997) with the removal rate after it adopts Daubert (post-1997). Thus, the before-after analysis measures change over time within the same group, presuming that any change in the tested variable is due solely to the treatment.\(^{211}\) The problem with the before-after analysis is that there may be some unknown covariate within the population that alters the measured variable and thereby produces potentially incorrect results.\(^{212}\)

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\(^{203}\) See supra Part II.D.1.e (discussing Cheng and Yoon’s errors).

\(^{204}\) See infra text accompanying notes 246–50.

\(^{205}\) See infra Part II.D.2.d.

\(^{206}\) Cheng & Yoon, supra note 6, at 485–86 n.43.


\(^{208}\) The “treatment” at issue is a state’s adoption of the Daubert standard.

\(^{209}\) The “tested variable” is rate of removal to federal court.

\(^{210}\) See LEE, supra note 207, at 64–65 (describing the before-after design).

\(^{211}\) See id. at 65.

\(^{212}\) See id. at 65, 99 (recognizing the danger posed by the “time effect” when using before-after analysis).
To avoid that problem, the difference-in-differences (DID) analysis combines before-after analysis with matching analysis.\footnote{See \textit{id.} (explaining how matching controls the undesirable effect that results from using a before-after design exclusively).} In matching, we compare two populations over the same period where one population receives the treatment and the other does not.\footnote{See \textit{id.} at 79, 99 (stating how matching allows researchers to measure two tested variables, where only one of the tested variables receives the "treatment" at issue).} Importantly, we must pick two populations that are comparable\footnote{See James J. Heckman et al., \textit{Matching as an Econometric Estimator}, 65 REV. ECON. STUD. 261, 261 (1998). Two populations are “comparable” if both populations “would have experienced the same outcomes . . . had they participated in the programme.” \textit{Id.} at 262. A central difficulty with matching is ensuring that the two groups are comparable. \textit{See id.} (explaining that it is difficult to determine whether two groups, having their “tested variable” measured, are truly comparable).} and are presumed to have the same covariate values.\footnote{See \textit{Lee}, supra note 207, at 99 (explaining that, if the effect of the treatment takes a long time to manifest itself, the changes observed between two control groups may be due to changes in other variables). If both groups have similar covariate variables, it becomes more likely that the observed differences between the two groups are due to the treatment. \textit{See id.}} In the Preliminary Study, Cheng and Yoon carefully argued that Connecticut and the EDNY are comparable in this way.\footnote{See Cheng & Yoon, supra note 6, at 486–87 (explaining the choice of Connecticut and the EDNY).}

By adding matching to before-after analysis, we can eliminate the effects of any unseen covariate in the treated population.\footnote{See \textit{Lee}, supra note 207, at 99.} In essence, if an unknown covariate is acting to decrease the rate of removals in the treated state (Connecticut), that same covariate should be acting in the same manner in the comparable state (New York).

The process for performing a DID analysis is very simple. First, calculate the mean rate of removal for Connecticut and the EDNY, separately, during the pre-treatment period (pre-1997) and the post-treatment period (post-1997).\footnote{The difference-in-differences method starts by selecting two tested variables with similar environmental conditions. \textit{Id.} Once the tested variables are chosen, the researcher must compare the subjects’ response variables before and after one of the variables received the treatment. \textit{Id.}} This provides four mean removal rates: (1) Connecticut pre-1997; (2) Connecticut post-1997; (3) EDNY pre-1997; and (4) EDNY post-1997. Next, calculate the difference within each time period between the treatment group and the non-treatment group.\footnote{See \textit{Joshua D. Angrist & Jorn-Steffen Pischke, Mostly Harmless Econometrics: An Empiricist’s Companion} 227, 229–30 (2009) (discussing the requirements of a fixed-effects difference-in-differences analysis); see also \textit{Lee}, supra note 207, at 99 (noting that the subjects’ response variables must be measured before and after treatment).} Thus, we calculate the difference between Connecticut pre-1997 and New York pre-1997, and between Connecticut post-1997 and New York post-1997. This results in the

\textit{Id.} at 262. A central difficulty with matching is ensuring that the two groups are comparable. \textit{See id.}
“differences” in difference-in-differences. Finally, calculate the difference between the post-treatment period differences and the pre-treatment period differences. For instance, in this case, take the post-1997 result and subtract it from the pre-1997 result. This results in the “difference” in the difference-in-differences.

b. A Mathematically Formal Description of the Difference-in-Differences Analysis

Intuitively, the idea of the DID analysis is easy to grasp, but examining the method in a formal mathematical notation reveals the kind of phenomena the DID approach is meant to analyze. To mathematically explain the DID analysis, we describe an algebraic method for doing a DID analysis that corresponds to the intuitive method. Then, we describe a method for doing the DID analysis using a fixed-effects analysis.

The algebraic explanation necessarily begins with the introduction of mathematical notation.

Let:

\[ \mu_{i,t} = \text{the mean of the study variable} \]

i = \begin{cases} 0 & \text{for the non-treatment group}\, 225 \\
1 & \text{for the treatment group}\, 226 \end{cases}

---

221. See ANGRIST & PISCHKE, supra note 220, at 230 tbl. 5.2.3 (illustrating how to calculate the difference-in-differences using the average employment in fast-food chains as an example); see also LEE, supra note 207, at 100 (discussing how the difference-in-differences rate was calculated to measure the effect immigration has on unemployment).


223. See ANGRIST & PISCHKE, supra note 220, at 233–41 (discussing difference-in-differences as fixed effects); see also MARNO VERBEEK, A GUIDE TO MODERN ECONOMETRICS 345 (2d ed. 2004) (noting that “[t]he fixed effects model is simply a linear regression model in which the intercept terms vary over the individual units . . . . We can write this in the usual regression framework by including a dummy variable for each unit i in the model”); Marianne Bertrand et al., How Much Should We Trust Differences-in-Differences Estimates?, 119 Q. J. ECON. 249, 250–51 (2004) (discussing difference-in-differences estimation and the fixed-effects regression formula); Card & Krueger, supra note 222, at 779–81 (utilizing a least-squares dummy variable model to discuss the effect of minimum wage changes on employment).

224. In this case, the study variable is the average removal rate.

225. In this case, the non-treatment group is the EDNY.

\( t = 0 \) for the pre-treatment period\(^{227}\) and \( t = 1 \) for the post-treatment period.\(^{228}\)

A DID analysis is performed by calculating the study variable mean for each population at specified time periods.\(^{229}\) In the Preliminary Study, the treatment group is Connecticut, the non-treatment group is the EDNY, the pre-treatment period is before 1997, and the post-treatment group is after 1997. Excluding 1997, the year Connecticut adopted \textit{Daubert}, in terms of the variable \( \mu \) would give us:\(^{230}\)

\[
\begin{align*}
\mu_{0,o} &= \text{EDNY from 1994 to 1996} = 0.5771 \\
\mu_{0,1} &= \text{EDNY from 1998 to 2000} = 0.7094 \\
\mu_{1,o} &= \text{Connecticut from 1994 to 1996} = 0.3259 \\
\mu_{1,1} &= \text{Connecticut from 1998 to 2000} = 0.3183
\end{align*}
\]

The difference-in-differences estimator is found by calculating the difference within each time period between the treatment group and the non-treatment group:\(^{231}\)

\[
\begin{align*}
\mu_{1,o} - \mu_{0,o} &= 0.3259 - 0.5771 = -0.2510 \\
\mu_{1,1} - \mu_{0,1} &= 0.3183 - 0.7069 = -0.3910
\end{align*}
\]

Finally, we calculate the difference between the post-treatment period differences and the pre-treatment period differences:\(^{232}\)

\[
\left( \mu_{1,1} - \mu_{0,1} \right) - \left( \mu_{1,o} - \mu_{0,o} \right) = -0.3910 - (-0.2510) = -0.1400.
\]

This gives us a difference-in-differences estimator of \(-0.14\%\) (or \(-0.0014\)), which means that Connecticut’s adoption of \textit{Daubert} decreased the removal rate by \(0.14\%\).

This number, standing alone, is not entirely helpful. It is important to know if it captures a statistically significant correlation between removal rate and

\(\footnote{227}{The pre-treatment period is pre-1997 because Connecticut adopted \textit{Daubert} in 1997. \textit{Id.}}\)

\(\footnote{228}{The post-treatment period is post-1997.}\)

\(\footnote{229}{See supra note 219 and accompanying text (outlining the periods that our study needs to break the study variables into in order to perform a difference-in-differences analysis).}\)

\(\footnote{230}{All values are shown in percentages. We ignore the year of transition throughout the analyses because it is highly likely that the data will be confused, owing a part of the year to the \textit{Frye} standard and a part to the \textit{Daubert} standard. Moreover, unless stated otherwise, the numbers used are based on our aggregation of the data, not Cheng and Yoon’s.}\)

\(\footnote{231}{See supra note 220 and accompanying text.}\)

\(\footnote{232}{See supra note 221 and accompanying text.}\)
adoption of *Daubert*. This question is answered by recognizing that the DID approach is simply a special case of the fixed-effects method used by Cheng and Yoon in the National Study.  

**c. Difference-in-Differences as Fixed-Effects Analysis**

Outside the context of a fixed-effects analysis, we would attempt, statistically, to capture the relationship between a state’s adoption of *Daubert* and removal rate by identifying a state where *Daubert* was adopted and then regressing the removal rate against years. If there was an effect of adopting *Daubert* on removal rate, then, after the first year of adoption, the rate of removal would change in a (hopefully) statistically significant way. Our discussion of the DID analysis should make us concerned that such a simple regression might fail because it does not take into account unknown covariates that could influence the removal rate in an unmeasured way and thereby interfere with finding a correlation between removal rate and the year. This problem arises because the study populations are likely to be heterogeneous.

In this case, heterogeneity may be apparent at both the state and year level. In other words, despite the assumption of comparability between the EDNY and Connecticut, each state might have its own unique set of characteristics that affect the way *Daubert* is correlated with removal rate. Similarly, each year might have its own unique characteristics that could affect the way *Daubert* is correlated with removal rates.

One way to capture this heterogeneity is to create a regression formula that contains a set of dummy variables to capture the state-specific and year-specific heterogeneity:

$$\text{Rate}_{it} = \alpha_1 + \beta_1 S_{1i} + \beta_2 S_{2i} + \gamma_1 Y_{1t} + \gamma_2 Y_{2t} + \ldots + \gamma_6 Y_{6t} + \delta D_{1t} + \mu_{it}$$

**EQUATION 2**

---

233. See supra note 222 and accompanying text.

234. See Heckman et al., supra note 215, at 262 (explaining how difficult it is to tell whether two groups are homogeneous). Ideally, if comparing data from two different subjects, we would prefer to assume that the statistical properties of the data for each subject are overall the same for the other subject. When this occurs, we have homogeneity. When this does not occur, we have heterogeneity; there may be unobserved relevant variables that are correlated with the observed variables and the value of those variables may differ from subject to subject. See Lee, supra note 207, at 64–65, 99 (discussing how changes that take place over time make it difficult to separate the “treatment effect from the ‘time effect’”). A heterogeneous population has unique or individual characteristics as compared to another population. See Damodar Gujarati, *Econometrics by Example* 282 (2011).

235. Dummy variables are variables that have a value of 1 if a condition is met and a value of 0 otherwise. Gujarati, supra note 234, at 47.

236. See id. at 283 (noting that equation 17.2 is the operation form of the “fixed effects regression model,” which accounts for heterogeneity).
Where the subscript \(i\) identifies the state \((i = 1\) when the state is Connecticut and 2 if the state is New York); the subscript \(t\) identifies the year \((t = 1\) for 1994, 2 for 1995, \ldots 7 for 2000),

Rate\(_{it}\) is the removal rate at time \(t\) in state \(i\);

\(S\)\(_{ni}\) is a set of two binary variables such that \(S\)\(_{1i}\) is set to 1 when \(i\) is 1 (the state is Connecticut) and to 0 when \(i\) is 2 (the state is New York), and \(S\)\(_{2i}\) is set to 0 when \(i\) is 1 (the state is Connecticut) and to 1 when \(i\) is 2 (the state is New York);

\(Y\)\(_{nt}\) is a set of six binary variables each of which is set to 1 when \(n\) equals \(t\);

\(D\)\(_{it}\) is a binary variable that is set to 1 if state \(i\) at time \(t\) has adopted Daubert and 0 otherwise;

\(\alpha\)\(_i\) is the y intercept;

\(\beta\)\(_i\) is the regression coefficient for state \(i\);

\(\gamma\)\(_i\) is the regression coefficient for year \(i\);

\(\delta\) is the regression coefficient for the Daubert variable and corresponds to the difference-in-differences estimator; and

\(\mu\)\(_{it}\) is the standard error term.

At this point, any heterogeneity in the state or the year should be captured by the model. The equation can be solved through a regression analysis by using removal rates. This method produces data that appears as follows:

\[\text{At this point, any heterogeneity in the state or the year should be captured by the model. The equation can be solved through a regression analysis by using removal rates. This method produces data that appears as follows:}\]

---

237. See Cheng & Yoon, supra note 6, at 489 (excluding 1997, the year of treatment).
238. \(Y\)\(_{1t}\) =1 when \(t = 1\), which corresponds to 1994, and \(Y\)\(_{6t}\) = 1 when \(t = 6\), which corresponds to 2000.
239. This depiction of the data may be somewhat misleading because, generally, we do not use all of the dummy variables in a fixed-effects analysis because of the “dummy variable trap.” See Gujarati, supra note 234, at 283. To calculate our Daubert estimator, we used a least-squares dummy variable model and performed a linear regression on that model. Because there was one dummy variable for each state and one for each year, we encountered perfect collinearity between one state dummy variable and one year dummy variable and the rate of removal intercept. See id. at 48, 283 (explaining how a perfect collinearity will lead to a common interval of 1). To avoid this problem, we dropped one state dummy variable and one year dummy variable. Id.
240. \(S\)\(_1\) is a dummy variable for Connecticut, \(S\)\(_2\) is a dummy variable for the EDNY, \(Y\)\(_1\) to \(Y\)\(_3\) are dummy variables for 1994 to 1996 respectively, \(Y\)\(_4\) to \(Y\)\(_6\) are dummy variables for 1998 to 2000 respectively, and \(D\) is a dummy variable for state adoption of Daubert. The year 1997 was dropped because it was the transition year.
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</table>

**Table 9**

The linear regression,\(^{241}\) using Equation 2\(^{242}\) and this dataset, produces a value for the *Daubert* coefficient $\delta$ of $-0.0013991$, which is extremely close to the result calculated using the algebraic method (-0.001400). Consistent with Cheng and Yoon’s conclusion,\(^{243}\) this result is statistically insignificant with a $p$-value of 0.329 and a 95% confidence interval of -0.0048985 and 0.0021003.

At this point, given the lack of statistical significance, Cheng and Yoon concluded that there was no measurable effect.\(^{244}\) That conclusion is incorrect. Evidence of statistical insignificance does not mean the underlying effect is zero; rather, it means that the data and the analysis provide insufficient evidence to determine the effect.\(^{245}\) If, instead of using a pure rates dataset, we

\(^{241}\) In this instance we used the same linear regression as Cheng and Yoon because it was consistent with our dual goals of explaining the statistical methods and demonstrating the problems with Cheng and Yoon’s methodology.

\(^{242}\) Importantly, the regression analysis will result in some of the dummy variables being omitted to avoid the problem of collinearity. See Gujarati, supra note 234, at 48, 283. In this case, the dummy variables for Connecticut and 1994 were omitted.

\(^{243}\) Interestingly, Cheng and Yoon do not provide any of the results from their DID analysis—neither the value of the *Daubert* coefficient nor the $p$-value. Instead, they merely conclude that the result is not statistically significant. See Cheng & Yoon, supra note 6, at 489–90.

\(^{244}\) See id.

\(^{245}\) See Siu L. Chow, Statistical Significance: Rationale, Validity and Utility 8 (1996) (noting that “an important real-life effect may be ignored simply because it does not reach
create a dataset that has one entry for each case filed in federal court, with each entry identified by state, year, and whether or not it was removed.\(^\text{246}\) We find a statistically significant correlation, at the \(p \leq .05\) level, between the adoption of \textit{Daubert} and removal rate: \(\delta = -0.0012657\) with a \(p\)-value of 0.006. This result is well within the 95% confidence interval calculated both in Cheng and Yoon’s linear regression on rates and in our own. Thus, the fact that regression using rates proves to be statistically insignificant does not, by itself, prove that there was no effect. Rather, it demonstrates a failure of the methodology to find the effect.

d. Using the Proper Study Population, the Preliminary Study Demonstrates That Connecticut’s Adoption of \textit{Daubert} Decreased Removal Rates

Using the proper removal rates, number of torts, number of removals, and an expanded dataset (the right study population), we were able to find that adoption of \textit{Daubert} by Connecticut in 1997 decreased the removal rates in a statistically significant manner.\(^\text{247}\) Further, our result is compatible with the 95% confidence interval provided by Cheng and Yoon.\(^\text{248}\) Thus, we can conclude that Cheng and Yoon were incorrect in their conclusion that \textit{Daubert} cannot be shown to have an effect on removals.\(^\text{249}\) Instead, the opposite is true—\textit{Daubert} does have an effect on removal rates and that effect is precisely what is expected if it is the stricter standard.\(^\text{250}\)

3. Recreating the National Study: Statistical Insignificance Caused by Inclusion of Confounding Data and a Small Effect Hidden by the Kind of Study Population Used

Having failed to find a statistically significant result in the Preliminary Study, Cheng and Yoon turned to the National Study where they used a fixed effects analysis on 16 states over 7 years to determine whether there was a statistically significant correlation between removal rate and state adoption of \textit{Daubert}.\(^\text{251}\) Here, we will also be unable to find a statistically significant
effect. But this failure is instructive because it unveils another error in Cheng and Yoon’s analysis: a failure to account for the fact that state adoption of Daubert before federal adoption differs in effect from state adoption of Daubert after federal adoption, leading to the inclusion of data from states that are not comparable.\textsuperscript{252} Once that error is eliminated, we find a statistically significant correlation between state adoption of Daubert and removal rate.\textsuperscript{253}

Similar to the fixed-effects version of the Preliminary Study, to engage in a fixed-effects analysis we must create two sets of dummy variables: one for the states and one for the years.\textsuperscript{254} Because we have 16 states, we will need 16 dummy variables:

\begin{align*}
S_{1,i} & \text{ is 1 when } i = 1 \text{ (the state is Alaska) and 0 otherwise;} \\
S_{2,i} & \text{ is 1 when } i = 2 \text{ (the state is Arkansas) and 0 otherwise;} \\
S_{3,i} & \text{ is 1 when } i = 3 \text{ (the state is Arizona) and 0 otherwise;} \\
S_{4,i} & \text{ is 1 when } i = 4 \text{ (the state is Connecticut) and 0 otherwise;} \\
S_{5,i} & \text{ is 1 when } i = 5 \text{ (the state is Florida) and 0 otherwise;} \\
S_{6,i} & \text{ is 1 when } i = 6 \text{ (the state is Indiana) and 0 otherwise;} \\
S_{7,i} & \text{ is 1 when } i = 7 \text{ (the state is Kansas) and 0 otherwise;} \\
S_{8,i} & \text{ is 1 when } i = 8 \text{ (the state is Michigan) and 0 otherwise;} \\
S_{9,i} & \text{ is 1 when } i = 9 \text{ (the state is Minnesota) and 0 otherwise;} \\
S_{10,i} & \text{ is 1 when } i = 10 \text{ (the state is Missouri) and 0 otherwise;} \\
S_{11,i} & \text{ is 1 when } i = 11 \text{ (the state is New Mexico) and 0 otherwise;} \\
S_{12,i} & \text{ is 1 when } i = 12 \text{ (the state is New York) and 0 otherwise;} \\
S_{13,i} & \text{ is 1 when } i = 13 \text{ (the state is North Carolina) and 0 otherwise;} \\
S_{14,i} & \text{ is 1 when } i = 14 \text{ (the state is Oregon) and 0 otherwise;} \\
S_{15,i} & \text{ is 1 when } i = 15 \text{ (the state is Tennessee) and 0 otherwise; and} \\
S_{16,i} & \text{ is 1 when } i = 16 \text{ (the state is Washington) and 0 otherwise.} \\
\end{align*}

We will also need 7 year variables:

\begin{align*}
Y_{1,t} & \text{ is 1 when } t = 1 \text{ (the year is 1994) and 0 otherwise;} \\
Y_{2,t} & \text{ is 1 when } t = 2 \text{ (the year is 1995) and 0 otherwise;} \\
Y_{3,t} & \text{ is 1 when } t = 3 \text{ (the year is 1996) and 0 otherwise;} \\
Y_{4,t} & \text{ is 1 when } t = 4 \text{ (the year is 1997) and 0 otherwise;} \\
\end{align*}

\textsuperscript{252} Cheng and Yoon included two states, Oregon and Arkansas, that adopted a liberal standard for admitting expert testimony similar to Daubert years before Daubert was decided by the Supreme Court. See id. at 493 fig.7.

\textsuperscript{253} See infra note 277 and accompanying text; see also infra FIGURES 6 & 7.

\textsuperscript{254} The inclusion of these dummy variables will help capture the heterogeneity between the study groups used in the data. See supra Part II.D.2.d.
\( Y_{5,t} \) is 1 when \( t = 5 \) (the year is 1998) and 0 otherwise;
\( Y_{6,t} \) is 1 when \( t = 6 \) (the year is 1999) and 0 otherwise;
\( Y_{7,t} \) is 1 when \( t = 7 \) (the year is 2000) and 0 otherwise;
This produces the following formula:

\[
\text{Rate}_{it} = \alpha_1 + \beta_1 S_{1,i} + \beta_2 S_{2,i} + \ldots + \beta_{16} S_{16,i} + \gamma_1 Y_{1,t} + \gamma_2 Y_{2,t} + \ldots + \gamma_7 Y_{7,t} + \delta D_{it} + \mu_{it}
\]

**EQUATION 3**

Running a fixed-effects analysis with Equation 3, using rates as the dependent variable, and using linear regression, the Daubert coefficient \( \delta \) has a value of -0.0008558 with a \( p \)-value of 0.871. Therefore, it is not statistically significant at the \( p \leq 0.05 \) level. In addition, unlike the Preliminary Study, changing the dataset to a one-entry-per-case dataset does not lead to a statistically significant result.

This result runs precisely counter to the result we achieve in the Flight-From-Frye thought experiment where we found a statistically significant result. To explain this difference, we recognize two differences that prevent Cheng and Yoon from identifying the true effect of the states’ adoption of Daubert on removal rates.

The first difference is that of size. We utilized a one-entry-per-case dataset and restricted the analysis to a pure Federal Database. Cheng and Yoon used removal rates based on the number of torts filed in state court. Thus,

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255. To avoid problems of collinearity, the regression eliminates one state dummy variable and one year dummy variable. See Gujarati, supra note 234, at 283. In this case, the state dummy variable for Alaska and the year dummy variable for 1994 were eliminated.

256. Our result differs from Cheng and Yoon’s because their calculation of removal rates in their National Study cannot be reproduced in a systematic way. See supra Part II.D.1.e. Our result is based on a calculation of the removal rate by limiting the set of cases to torts, from 1990 to 2000, that were removed to federal court, and under diversity jurisdiction. According to Cheng and Yoon, \( \delta \) has a rounded value of -0.0053 with a \( p \)-value of 0.951. Cheng & Yoon, supra note 6, at 497.

257. See infra notes 258–77 and accompanying text. These errors are in addition to the difficulties created by Cheng and Yoon’s errors in calculating rates of removal. See supra Part II.D.1.e.

258. Thus, we calculated removal rates as the number of cases removed to federal court divided by number of torts filed in federal court. This removal rate was then used to create a dataset that had, for each state and year, as many entries as torts filed in the federal court of that state and year and where each entry was identified (in a ratio consistent with the removal rate) as either removed or not removed from state court.

259. See Cheng & Yoon, supra note 6, at 491–92 (discussing how Cheng and Yoon gathered state court data regarding the number of torts filed in state court to calculate removal rates in their National Study).
our database is both larger and smaller than Cheng and Yoon’s. It is larger because we have 57,640 entries, while Cheng and Yoon have only 112 (a difference of 514.6-to-1). But, it is also smaller because Cheng and Yoon’s removal rates are calculated with a denominator based on tort filings in state court. Thus, their removal rates summarize 2,956,439 cases while ours correspond to only 57,640 (a difference of 51.3-to-1). Both of these size differences work to our advantage and Cheng and Yoon’s disadvantage.

The large size of our federal dataset compared to Cheng and Yoon’s plays an important role. Assuming the Daubert effect is small, it is more likely that a statistically significant result will be found in the larger dataset. This is because our 57,640 data entries enable us to more easily identify a statistically significant result. Similarly, Cheng and Yoon summarized a vastly larger dataset which will likely introduce many torts for which Daubert is not relevant. Once again, if we assume that the Daubert effect is small, then it will be harder to find the effect if the data includes cases for which Daubert adoption is irrelevant because those cases will create a noise—masking the effect. Thus, our 57,640 data entries work to our advantage.

The second difference is the set of state Cheng and Yoon examined. They analyzed 16 states while we look only at the 13 states that adopted Daubert after the federal courts. This difference matters because, if defense attorneys see Daubert as a stricter standard, then those attorneys will tend to stay in state court if they are in a state that adopted Daubert before the federal courts to

260. See infra notes 261–62 and accompanying text.
261. See Cheng & Yoon, supra note 6, at 497, 500–02 & fig.9 (showing that Cheng and Yoon had 110 entries after the 2 dummy variables were removed to avoid collinearity).
262. These numbers were calculated using Cheng and Yoon’s data and reviewing the number of cases filed per state and year.
263. See infra notes 264–69 and accompanying text (explaining our method used).
264. See Chow, supra note 245, at 90–91 (discussing the sample-size dependence problem: whether or not statistical significance is obtained often depends on the sample size of the data).
265. Cf. id. (discussing how statistical significance is easier to find when there are larger numbers of entries tested).
266. See Cheng & Yoon, supra note 6, at 491–93; see also supra note 262 and accompanying text.
267. Cf. Gross, supra note 11, at 1119 (noting that there are certain torts in which the rate of expert witness testimony is much higher).
268. This could be a covariate in the data because distinct states are likely to have a separate ratio of tort cases based on their location. See LEE, supra note 207, at 64–65, 99 (discussing the problems covariates cause when comparing tested variables that take a long time to feel the effects stemming from a “treatment”).
269. Cheng & Yoon, supra note 6, at 493.
270. Arkansas and Oregon adopted tests similar to Daubert as a state evidentiary standard before the federal courts adopted Daubert in 1993, and New Mexico adopted the standard in the same year. See State v. Alberico, 861 P.2d 193, 203–04 (N.M. 1993); see also Prater v. State, 820 S.W.2d 429, 431 (Ark. 1991); State v. Brown, 687 P.2d 751, 759 (Or. 1984); supra note 139.
gain the evidentiary advantage in state court.\footnote{271} Once the federal courts adopt \textit{Daubert}, there is a leveling of evidentiary standards.\footnote{272} We would expect this to result in an increase in the rate of removals because defense attorneys would no longer use the evidentiary difference as a basis for staying in state court.\footnote{273} This is precisely what we find.\footnote{274}

We engaged in two sets of logistic-fixed-effects analyses from 1990 to 1996,\footnote{275} one on pure federal data and the other using mixed federal and state data, using 10 states: Arizona, Arkansas, Florida, Kansas, Michigan, Minnesota, Missouri, New York, Oregon, and Washington. Arkansas and Oregon both adopted a \textit{Daubert} standard before 1993 while the 8 other states did not adopt \textit{Daubert} during the study period (and stand in as controls in our analysis).\footnote{276}

Both sets of analyses yielded statistically significant results at the $p \leq .05$ level,\footnote{277} demonstrating that the probability of removal increased after the federal courts adopted \textit{Daubert}:

<table>
<thead>
<tr>
<th>\textit{Daubert} Adopted By Federal Courts</th>
<th>Probability of Removal</th>
</tr>
</thead>
<tbody>
<tr>
<td>No</td>
<td>0.287348</td>
</tr>
<tr>
<td>Yes</td>
<td>0.330554</td>
</tr>
</tbody>
</table>

\textbf{FIGURE 6: FIXED-EFFECTS ANALYSIS ON REMOVAL RATE USING FEDERAL DATA ONLY}

\footnote{271} See Horowitz, \textit{supra} note 113, at 18 (arguing that defense attorneys should remove the case when it is in their clients’ best interest); \textit{see also} Miller, \textit{supra} note 113, at 24, 28 (arguing that a defense attorney can win a case with a successful \textit{Daubert} motion).

\footnote{272} There is a leveling of standards because, in our hypothetical, both the state and federal courts apply the same evidentiary standard: \textit{Daubert}.

\footnote{273} \textit{See supra} note 271 and accompanying text.

\footnote{274} \textit{See infra} note 277 and accompanying text; \textit{see also supra} \textit{FIGURES} 6 & 7 (showing the results of the calculated removal rates).

\footnote{275} We excluded 1993 because that is the year the federal courts adopted \textit{Daubert}. \textit{Daubert} v. Merrell Dow Pharms., Inc., 509 U.S. 579 (1993).

\footnote{276} \textit{See supra} notes 140, 270 and accompanying text. We eliminated New Mexico, which adopted \textit{Daubert} in 1993, although it is unlikely that the adoption of \textit{Daubert} by the federal courts would have any effect on removal rates in New Mexico because, before 1993, the federal and state courts used the same evidentiary standard, and, in this case, even after 1993, they both used the same standard. \textit{See State v. Alberico}, 861 P.3d 192, 203–04 (N.M., 1993). Thus, its inclusion in the analysis would only create the same error found in Cheng and Yoon’s analysis.

\footnote{277} The $p$-value for both logistic analyses was $< 0.0005$. Both analyses also produced statistically significant results using linear regression showing that the correlation was positive between federal court adoption of \textit{Daubert} and removal rate.
FIGURE 7: FIXED-EFFECTS ANALYSIS ON REMOVAL RATE USING FEDERAL AND
STATE DATA

Cheng and Yoon’s analysis uses one variable, the Daubert dummy variable, to account for two very different effects: (1) the effect of state adoption of Daubert where the state adopts after federal adoption and (2) the effect of state adoption where the state adopts Daubert before federal adoption.278 Because the two different effects work in opposite directions (increasing removal rate and decreasing removal rate, respectively), combining them is likely to result in one cancelling the other out and making it nearly impossible to find any statistically significant effect.279

III. DAUBERT IS THE STRICTER STANDARD FOR EXPERT WITNESS
ADMISSIBILITY

A. Implications of the Daubert Effect Conclusions

Our analysis demonstrates that the choice between Frye and Daubert does matter, and matters quite significantly.280 Courts addressing the substantive scientific admissibility standard must consider the effect of choosing between two different substantive standards.281 If a court chooses Daubert, the ramifications will be that litigants will perceive the standard as strictly assessing expert evidence.282

The difference in scientific admissibility standards also has considerable effects on substantive tort law.283 Of course, the burden of proof in civil cases generally falls on the claimant, usually by a preponderance of the evidence standard.284 But, if the claimant must also prove issues that require expert

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278. See supra notes 269–76 and accompanying text.
279. See supra notes 132, 277 and accompanying text (discussing the statistically significant results found in our study).
280. See supra notes 133, 141, 277 and accompanying text.
281. Cf. Krafka et al., supra note 4, at 22–23 (explaining how judges’ actions in the post-Daubert era are indicative of the fact that the change in evidentiary standards has affected how they approach scientific evidence).
282. See id. at 22–24.
283. See infra notes 284–91 and accompanying text (expressing the significance of an expert testimony admissibility standard).
284. See Stein, supra note 14, at 81 (“For civil trials, the default proof standard is a preponderance of the evidence. Under this standard, adjudicators must deem a factual allegation
testimony, the shift to Daubert will raise the requirements for the claimant to prove his or her case. Whether this shift is good or appropriate is not part of our analysis, except that any choice must be done knowing the full effect of the decision.

Further, our study raises an issue that relates back to the origins of the Daubert standard itself. In Daubert, Justice Blackmun, writing for the majority, rejected the Frye standard because it was so inconsistent with “the liberal thrust of the Federal Rules.” Despite this proclamation, we now see that the Daubert standard has tightened scrutiny beyond the Frye limits that troubled Blackmun and the majority. If so, the Court badly misjudged the effect Daubert would have on expert testimony, and the Court must now revisit the issue to clarify the standard, either affirming the new reality or reanimating the lenient standard originally intended.

Lastly, the results of this study also have an effect on litigants in court. The effect of stricter scrutiny will change the marketplace for expert services because only those experts that can provide a sound empirical basis for their opinions will be admitted. In a world of strict expert review, empirical testing will be the lynchpin of admissibility. Therefore, litigants encourage and support specific fields when they choose to hire fields that satisfy the criteria of empirical science, while other fields will be forced to adapt to the new reality or fade away.

B. Potential Areas of Future Research

There are several areas of further research that would build upon our analysis.

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285. See supra Part II.D.2.d (explaining that Daubert is a stricter standard on litigants because it results in a higher rate of removals).


287. See supra Part II.E (concluding, instead, that Daubert has resulted in a stricter standard).

288. See Faigman et al., supra note 29, at 1810 & n.34 (describing the widely shared view that the Frye standard would prove to be a stricter standard for admissibility than Daubert as “an erroneous one”).

289. See Saks, supra note 10, at 239 (“Testimony from those fields that could, but do not, provide adequate data . . . should be excluded until they can provide adequate empirical support for what the experts are claiming for their field in general, or for themselves individually.”).

290. See id. at 238–39 (“For those fields that can answer essential questions about themselves with sufficient completeness . . . the decision whether to admit or exclude can be made with relative ease.”).

291. See id. at 240 (observing that Daubert created a difficult situation for sciences that cannot satisfy the empirical data criteria).
similar, set of data. As mentioned previously, the success of the analysis necessarily depends on the underlying data quality. The analysis here relies, in whole or in part, on the federal dataset. The same analysis, utilizing exclusively state-level data, may be useful to confirm or rebut these findings.

A second area of further research evolves from a limitation of this study. In removal rate analyses, the choice for removal lies with the defendant. Although the removal rate analysis is a powerful tool to measure these litigants’ behavior, it can only capture about half of civil case actors. The other half, the plaintiffs/claimants, must be studied using other metrics. Use of these alternative metrics would result in a more complete picture of Daubert’s effects.

A third potential study builds on the innovation of aggregate case data in a more general sense. The removal rate metric is capable of capturing a considerably larger number of cases than a case study approach. For example, this study used data on nearly 4 million cases, as compared to the several hundred used in other studies. Just as our study applies the removal rate metric to civil case data, expanding on case review studies such as Dixon and Gill, a researcher could apply an aggregate case data metric to criminal case data as well, expanding on the case review approach of the Grosocup study. This type of study could measure the aggregate effect of Daubert on criminal cases and compare it to the effect on civil litigation. Additionally, aggregate case data methodologies can thoughtfully be applied to other substantive areas beyond Daubert.

Finally, because Daubert is a more restrictive test, one might start to wonder if it has become synonymous with, or at least subsumed within the sphere of,

292. See supra Part II.D.1 (arguing that working with data that is fundamentally flawed will lead to inaccurate conclusions).
293. See supra Part II.C.2 (relying on the federal dataset).
294. See supra note 88 and accompanying text (describing the process for removal to federal court).
295. See supra text accompanying note 294 (noting that it is the defendant’s choice to remove to federal court and, consequently, this rate will only capture the defendant’s actions).
296. See supra notes 294–95 and accompanying text (noting the limitations of this study).
297. See infra note 298 and accompanying text (providing an example of this difference because the cited removal rate data approach was able to capture approximately 4 million cases when the cited case study approach was only able to capture 399).
298. See Dixon & Gill, supra note 4, at xiii (analyzing 399 district court opinions); see also supra note 106 and accompanying text.
299. Grosocup et al., supra note 3, at 344 (analyzing 693 appellate opinions).
300. Cheng and Yoon cite one example of aggregate case data methodologies being applied to an area other than to Daubert. For example, they credit a 2002 Federal Judicial Center study on class action lawsuits for the origin of removals as an aggregate measure of procedural change. See Cheng & Yoon, supra note 6, at 482 n.32 (citing Bob Niemic & Tom Willging, Fed. Judicial Ctr., Effects of Amchem/Oritz on the Filing of Federal Class Actions: Report to the Advisory Committee on Civil Rules 12–13 (2002)).
summary judgment. A very strict admissibility standard could have the same effect, in practice, as summary judgment by limiting the evidence available to claimants. *Daubert* may have not reached that level, but it is important to consider the interrelationship of the two issues.

IV. CONCLUSION

Since 1993, the effect of *Daubert* on expert admissibility has been hotly contested. In *Daubert*, the Supreme Court indicated strongly that this test would be more lenient for admissibility of expert evidence. Yet the initial perception of the effect was not overwhelmingly consistent with that result, so researchers began to investigate whether and how the scientific admissibility standard had changed. Some survey data indicated that judges saw *Daubert* as a stricter standard, while some did not. Some case review analysis found that *Daubert* was a stricter standard, while some did not. Despite utilizing several different empirical methodologies, the final determination of *Daubert*’s true effect remained inconclusive.

Finally, in 2005, Edward Cheng and Albert Yoon offered a fresh approach, utilizing removal rates to measure actual litigant behavior based on *Daubert*. Their approach captured several advantages over other approaches by: (1) aggregating large amounts of case data; (2) measuring real-world behavior without reporting a bias; and (3) isolating the scientific admissibility standard variable.

We also believe removal rate analysis is the best tool for assessing the aggregate effect of *Daubert*, so we recreated Cheng and Yoon’s underlying removal rate data to conduct additional studies. The results of this analysis demonstrated that, based on actual behavior in millions of real cases, civil defendants believe the *Daubert* standard to be a stricter one. Not only does the removal rate increase in the years after *Daubert*, as one would expect if the standard for admissibility is tighter, but we can also show that if the state adopts *Daubert*, and in so doing returns the state and federal court to the same admissibility standard, the removal rate then drops in response. Both of these effects support the conclusion that defendants perceive *Daubert* as an advantageous, stricter standard.

Because the results discussed in this Article so clearly contrast with Cheng and Yoon’s study, we then revisited the analysis Cheng and Yoon performed, looking for an explanation for the different results. When we did so, we found methodological problems with both the underlying data aggregation and the mathematical analysis of those data. Consequently, these errors undermine the validity of their study and their conclusions cannot be supported. When using our methods, the answer is clear: *Daubert* is the stricter standard of admissibility.