The America Invents Act and Innovation by Small Entities

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Abstract

This paper studies patenting activity by small U.S. entities before and after the passage of the Leahy-Smith America Invents Act (AIA) of 2011 which changed the patenting rule in the United States from a first-to-invent to a first-inventor-to-file system. Prior to the AIA, entities had the benefit of flexibility on when to file for patents but this benefit came at a cost; it created uncertainty in an atmosphere of litigative behavior about the date of the invention. To curb litigation, the AIA intended to reward disclosure of inventions through patent filing, but this added to the already constrained budget of small entities. Contrary to the AIA's intentions, the results indicate a decline in patenting activity post-AIA for all entities, and a widened gap between small and large entities compared to the pre-AIA period. Further, a higher exposure to litigation results in a decline in patents filed, but the rate of decline stalls post-AIA. This paper makes two contributions. First, it provides empirical evidence about the impact of AIA on small entities' patenting behavior and empirically tests if the concerns laid out by legislators prior to the AIA's enactment hold. Second, it lays out important differences between small and large entities' portfolio of patents and their incentive to file for patent rights.

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1 Introduction

This paper studies the America Invents Act (AIA) and its role in changing the way small entities file for patent rights in the United States. AIA is a complex legislation with multiple objectives and it significantly modified the U.S. patent rules. To this date, its impact on patenting behavior remains understudied. Small entities employ one-third of the U.S. labor force and innovative small entities are special for the U.S. economy as they take steep risks by investing their limited resources into the development of radical inventions. Theoretical predictions do not clearly indicate if the newly adopted rule in 2011 benefited all entities, and on the contrary, the scant evidence on AIA and similar legislations passed in other countries indicate its adverse effects on small entities' and inventors' patenting activity. This paper studies the primary aspect of the AIA, the move of the U.S. patenting system from a first-to-invent to a first-inventor-to-file rule, and determine empirically whether it changed the quantity and quality of patents filed by small entities compared to large entities at the United States Patent and Trademark Office (USPTO).

The Leahy-Smith America Invents Act (AIA) of 2011 substantially changed the patenting rules in the United States by adopting a first-inventor-to-file (FITF) rule. A FITF rule recognizes the *first filer* of an invention as the sole inventor, as opposed to a first-to-invent (FTI) rule, which recognizes the *first inventor*, regardless of the filing date. The overarching goal of the AIA was to encourage disclosure of inventions by rewarding the filing of patents at an early stage of the invention's development which would reduce uncertainty in its date of filing and would curb any litigation that arises due to an uncertain date of invention.[1] The previous rule, FTI, provided an inventor the flexibility of inventing first without the worry of rushing to file for patents. While this rule is beneficial to inventors, because they can focus solely on the invention, it also generates uncertainty

[1] An invention can be kept secret, or can be disclosed to the public, either freely or through patents. Disclosure proliferates follow-on invention. But, the definition of disclosure of an invention can vary with context, and so does its intensity. From only declaring the name of a new invention, to a few mentions on certain websites without its specifics to laying down the exact steps to recreate it, everything can be counted as its disclosure. See De Rassenfosse et al. (2020) and Rantanen (2012) for a discussion on disclosure, patents, and its relation to follow-on invention.

on when a patent is actually filed. This uncertainty is partially responsible for attracting patent lawsuits, and it is not conducive to innovation. Along with this, there were a few more secondary reasons that motivated the development of a patent reform act. But the sweeping and broad reform brought forth by the AIA — recognizing the first filer — also implicitly asks an inventor to rush to the patent office immediately after invention, and entities with constrained budgets, such as small entities, may not have the luxury of doing so.[2] Small entities behave differently, have markedly different patenting strategies compared to large entities, introduce disruptive inventions, drive innovation in new directions, and are litigated at a higher rate (Aghion et al., 2021; Abrams et al., 2019). It is unknown how they modified their portfolio of patenting activity (i.e. their quantity and quality) post-AIA. Patents' quality can be heterogeneous and changes as a strategic response to changes in incentives to patenting (Mezzanotti, 2021). This paper measures small entities' portfolio of patents — using the number of patents and their quality — using a measure of citation and compares them to large entities before and after the passage of this legislation using a difference-in-differences specification.

Answering these questions contributes to the following: first, provides evidence in support of the longstanding concern among the U.S. Congress members wary of the detrimental effects of AIA on small inventors. Studies indicated that the divide between large and small entities' (i.e. small businesses and individual inventors) patenting activity could widen after the AIA's passage and this paper confirms such predictions (Abrams and Wagner, 2013; Lo and Sutthiphisal, 2009). Second, it reiterates the challenges small entities always face while inventing and filing patents and highlights how the benefits and costs of disclosing inventions in the form of patents differ for entities of different sizes and change as a result of legislation-induced incentives to stimulate innovation. This paper also reiterates the fact that there are multiple objectives of the legislation that require serious attention and that this paper only studies one of the AIA's aspects. Studying the AIA is important as it will help design the next patent reform bill better.

[2] A small entity can be a person, a small business concern, or a nonprofit organization. If the entity is a small business, it is defined as having less than 500 employees. Universities and institutes of higher education are categorized under nonprofit organizations. See: https://mpep.uspto.gov/RDMS/MPEP/e8r9#/e8r9/d0e30961.html

Since the advent of the patenting system, the United States followed a different set of patenting rules, the FTI, in comparison with the rest of the major patent-granting countries, that followed the FITF. Both the set of rules have their benefits and costs. The understanding and assumptions that define such benefits and costs have varied from country to country. As a substantial part of these benefits and costs are immeasurable, there is no other way but to rely on country-specific case laws and legislation. As a result, the theoretical comparisons between FITF and FTI rules do not yield a clear winner as to whether following one is strictly better than the other; rather the comparisons suggest both can be conducive to innovation under given conditions (Miyagiwa, 2015; Scotchmer and Green, 1990). In addition to this, litigation is uncertain and too much of it can seriously undermine the validity of patents and as a result, discourage patenting. This also has a cascading effect on follow-on inventions (Mezzanotti, 2021; Kiebzak et al., 2016; Tucker, 2013; Lanjouw and Schankerman, 2001). On the flip side, a FITF rule limits a part of these litigations but comes at the cost of quickly filing for patents as soon as an invention is completed. Anecdotes from U.S. small entities suggest that they often invent first and then search to secure resources to file the invention as a patent. Resources are not only limited to funds but also include assistance from attorneys and agents.[3,4] Needless to say, these entities have to also worry about their idea being stolen, now more than before. U.S. Congress was therefore divided when the AIA was discussed. [5] The AIA aimed to curb the increase in litigation and also included provisions to limit the increase in costs of filing patents for smaller entities and in particular individual inventors.[6] Further, to study the effects of the AIA on small entities, a study was commissioned (Lerner et al., 2015). Due to a large number of pending patent applications which resulted in a severe

[3] https://www.nytimes.com/2012/02/09/business/smallbusiness/business-owners-adjusting-to-patent-system-overhaul.html

[4] In 2015, the cost of filing a patent in the United States ranged from $6000 for an "extremely simple" patent application, to more than $19,000, for a "highly complex" patent application. https://www.ipwatchdog.com/2015/04/04/the-cost-of-obtaining-a-patent-in-the-us/id=56485/

[5] Senator Feinstein's argument on the AIA being detrimental to small entities: "[t]his presents a particular hardship for independent inventors, for startups, and for small businesses, which do not have the resources and volume to employ in-house counsel but must instead rely on more-costly outside counsel to file their patents. This added cost and time directed to filing for ideas that are not productive will drain resources away from the viable ideas that can build a patent portfolio–and a business."

[6] The AIA created a new category of inventors, called "micro entity", who can avail 75 percent discount in all fees. This is discussed in detail in Section 2.1

lack of data at the time when the study started, the authors could not provide a definite answer as to whether the AIA benefited or harmed small entities' patenting activity.[7] Therefore, we still do not know if these provisions were enough to offset the increase in costs of patenting for small entities and provide a net benefit. Thus, this paper studies and provides evidence of changes in patenting activity before and after the AIA in the United States with a focus on small entities. Change in patenting activity can be considered the best signal to study the participation of entities of various sizes in the patenting process. Thus, in particular, the first question this paper asks is: was there a change in the number of patents filed by small entities relative to large entities post-AIA? Patent counts do not indicate if an invention is truly new, or if it is written clearly enough to be replicated by experts — all of which if true, can indicate a patent is of high quality (Hall et al., 2004). Therefore, I also ask a second question for a comprehensive understanding of the total effect of the AIA on the patenting activity of small entities: did the quality of patents filed by small entities of change post-AIA when compared to large entities? Studying the quantity and quality jointly can provide us with a better understanding of the total effect of AIA on small entities' patenting activity.

Early evidence indicates a reduction in disclosure of inventions by larger entities post-AIA; however, for smaller entities, the evidence remains inconclusive partly due to limitations in the data and partly due to unexplored and thus undefined scope of innovative activity in small entities (Lerner et al., 2015; Huang et al., 2020). Small entities have different incentive structures and innovative behavior. This is not represented adequately by publicly listed companies. Along with this, evidence from an AIA-like reform from Canada reported adverse effects on small firms and individual inventors upon its implementation (Abrams and Wagner, 2013; Lo and Sutthiphisal, 2009). This further warrants the identification of small entities and the study of AIA's effect on their patents.

To address these questions, I connect various datasets from the USPTO and add

[7] Two major limitations, as the authors note: "Only a small amount of data exists for patent activity under the FITF rules since (a) the FITF provision became effective on March 16, 2013 and (b) there was still, at that point, a major backlog of patents under the old system because the patenting process — from application to issue/abandonment — took close to 2.5 years for fiscal year 2013. The complexity of the law has led to varying interpretations from small businesses and small business investors, which has likely been reflected in varying responses."

entity-level disambiguated identifiers from the Patentsview database to construct an entity-quarter-level dataset. I use the number of patent applications applied by different entities as an indicator of invention disclosure, which is the broadest indicator of innovative activity at the entity level. The number of patent applications captures the resulting effect of all the policy changes the AIA introduced, and this approach to the data is in line with the literature. To measure the change in the quality of patents, I use the number of citations a patent received in its first two years of issue per patent and rescale it to a percentile value for its field of invention — also called average scaled citation. These two measures used for the main set of analyses are not the only measures of quantity and quality of innovation. In recent years, the value of innovation captured through changes in stock value for public entities as well as measures of quality developed from patent texts have been widely used as a proxy for patent value (Kogan et al., 2017; Kelly et al., 2018). This measure is unfortunately only available for publicly listed companies. All the measures highlight one particular dimension of innovation and none are adequate on their own. Therefore, I will show in this paper a range of alternative specifications and alternative measures in an effort to highlight the nuances of patent analysis and corroborate my findings.

I use a difference-in-differences approach to estimate the change in the number of patents and scaled citations for small entities relative to large entities post-AIA. I find that on average, small entities are filing for a lower number of patents post-AIA relative to large entities, but the average quality of the patents filed has not significantly changed post-AIA. This constitutes AIA's overall effect on the quantity and quality of patent applications and is this paper's primary set of results.

Next, I define a secondary specification to estimate the same outcomes — the change in the number of patents and scaled citations when entities are exposed to different levels of litigation. We know that certain fields of the invention are more prone to litigative behavior than others (Marco and Miller, 2019). This tests whether the AIA's objective to reduce litigation led to an increase in patenting activity for different entities especially small entities or not. According to the AIA's mechanism, I should observe an increase in confidence among inventors to publicly disclose their inventions now that litigation was supposed to drop. What I observe is over the years approaching the AIA's passage, the number of patents and scaled citations continued to fall in areas that have a larger exposure to litigation but with the AIA's passage, this fall was arrested. While this indicates a relative increase in confidence among small entities to file for patents post-AIA, this has to be investigated further as post-AIA, a landmark case — *Alice v. CLS Bank* was decided that changed the definition of eligibility of patents. This will be discussed in detail in Section 5.2. As robustness, I study patenting activity for a sub-group — publicly listed companies and I find no change in the outcomes for small publicly listed companies as compared to large ones post-AIA. This reiterates the fact that the publicly listed small entities are in reality too large and behave just as any other large publicly listed company and thus do not adequately represent a large chunk of small entities who file for patent applications.

The remainder of the paper is organized as follows: in Section 2 I detail the provisions of the AIA and compare it with other countries. In Section 3, I list the data sources, the sample selection, and the definition of the variables. Next, in Section 4, I explain the empirical strategy, and lastly, in Section 5, I discuss the inferences derived from the estimates. Finally, I conclude by summarizing my findings in Section 6.

2 The America Invents Act (AIA) of 2011

2.1 Provisions of the AIA

The U.S. patent regime underwent significant changes since World War II and the most recent amendment to the 35 U.S. Code—the America Invents Act of 2011 is regarded as the most substantial one since the Patent Act of 1952, changing the regime from First-to-Invent (FTI) to First-Inventor-to-file (FITF) (Lerner, 2000).[8] Under an FTI rule, an inventor had the option to claim that they are the original inventor even if another inventor had filed a patent application for a similar invention before theirs. Under a

[8]https://www.nytimes.com/2011/09/09/business/senate-approves-overhaul-of-patent-system.html

FITF rule, the inventor who files for patent rights first is the rightful owner of the patent. Pre-AIA, the invention date could be used to claim rights that do not matter post-AIA, and instead the date of filing matters (Masur and Ouellette, 2020). While the former rule provided flexibility to an inventor as to when to file for patent rights, entities could abuse this flexibility for strategic advantages. A group of entities, commonly known as "patent trolls" or Non-Performing Entities (NPEs) is those who do not invent; but rather acquire patents and assert their rights to invalidate other patents. Such entities wait for other inventors to file for a similar patent application as theirs or start producing a product that uses an invention similar to theirs. When the producing entities apply for patent rights or start production, the NPEs file for injunctive relief and ask for hefty royalty for infringement. This action is known as a hold-up. Due to the fear of hold-ups or infringing upon other patents, inventors restrict the disclosure of their inventions. As a result, the innovation of an economy does not reach its optimum level (Tucker, 2013).

The AIA's FITF rule tries to address the problem of unnecessary lawsuits partially. While not all kinds of patent lawsuits will be taken care of by the implementation of the AIA, at least for the cases related to the invention date, the AIA tries to establish a more certain regime by not providing the opportunity to contest conflicts on the date of invention. Pre-AIA, conflicts on the date of the invention were litigated under interference proceedings. An interference proceeding would be conducted when one patent application interferes with another in the process of its filing.[9] Post-AIA, this has been changed to derivation proceedings. Now, an entity cannot claim rights if they were late to the patent office. But, if they can show that an invention already in patent office was derived from their invention, they can file for a derivation proceeding.

[9]To understand interference proceedings and its relation to the AIA, I present an example. In a pre-AIA period, an invention i_1 came into existence at t_1 but was not filed at the patent office and an independent invention i_2, also in the pre-AIA period, was invented at t_2 ($t_2 > t_1$) and was filed as a patent application immediately. Here, invention i_2 interferes with invention i_1. Invention i_1 can still claim its patent rights or at the least negate i_2's patentability by conducting an interference proceeding, which will be conducted as a lawsuit. In the post-AIA period, in a similar situation, i_2's patent application will be upheld as the only patent application and i_1 can contest if it files for a derivation proceeding. Derivation proceeding replaced interference proceeding post-AIA and in the example, i_1 will claim that i_2 has been derived from i_1. But in the post-AIA period, invention i_2 will have stronger protection and i_1 will face higher costs because i_1 will have to show that i_2 has been derived from i_1 and not independently invented. A low count of interference proceedings does not mean that they are seldom used. Rather, it indicates that the patent applications prone to such opposition are deterred from filing.

To cushion small entities from bearing the extra cost of quick filing, the AIA implemented certain provisions. The AIA sets up two programs to assist small entities to file their patent applications, called the Pro Bono Program and the Patent Ombudsman for Small Businesses. The Pro Bono Program provides the qualifiers with free legal assistance in preparing and filing patent applications. [10] The Patent Ombudsman program assists applicants when a normal application stalls. [11] The AIA defines a new category of entities, called "micro-entity" for whom the USPTO levies a discount of 75 percent on all patent filing fees. [12]

The AIA not only changed the patent rules, but introduces post grant review (PGR) and inter partes review (IPR), replacing the ex parte reexamination. Any third-party can challenge the validity of a patent, using PGR in the first nine months, and then using IPR. As opposed to an ex parte reexamination, where a third party could apply for reexamination of a granted patent but not have active participation in the proceedings, PGR and IPR allows the third party to actively take part in the challenge process. Introduction of new and cheaper ways to oppose a patent's validity can caution entities who operate in areas predominantly exposed to litigation. To put numbers, a court litigation can run to millions of dollars while PGR/IPR can be estimated to be at around $500,000 (Masur and Ouellette, 2020).[13]

2.2 Comparison with other countries and the unintended effects of the AIA

Most countries have always followed a FITF rule. To date, three countries have made a move from FTI to FITF. Since 2007, different patent reform bills have been highlighting the need of harmonizing the patent system in the US with the rest of the world, and by the advent of AIA, the US is the third and last country to make such a change (Matal,

[10] https://www.uspto.gov/patents/basics/using-legal-services/pro-bono/patent-pro-bono-program

[11] https://www.uspto.gov/patents/ombudsman-program

[12] A micro-entity is a small entity with additional thresholds. A micro-entity has to qualify for an income threshold, which is three times the median income household. Note that institutes of higher education are considered micro-entities. The detailed definition can be found here: https://www.uspto.gov/patents/laws/micro-entity-status

[13] https://www.uspto.gov/patents/ptab/trials/post-grant-review

2011a; Matal, 2011b). In 1989 and 1998 Canada and the Philippines made a similar move respectively.[14,15] Only a few studies exist on the evaluation of patenting activity by entity types around the AIA and the Canadian reform and no study evaluates the Philippines reform. A common theme from all the studies points toward no significant benefit to smaller entities after such a change.

In Canada, patenting activity skewed towards large firms compared to small and independent inventors after their reform (Lo and Sutthiphisal, 2009). Additionally, a drop in the number of patents is reported among independent inventors with no change in their patent quality (Abrams and Wagner, 2013). In the US, studies find a decline in innovative activity, especially among publicly traded firms, after the AIA. Huang et al. (2020) use narrative R&D disclosure of publicly traded firms to measure innovative activity and find a decline, especially for the innovative firms among all publicly traded post-AIA. A narrative R&D is different from the R&D expenditure reported by firms in their annual reports. The authors use the number of R&D-related sentences counted from the firms' 10-K filings. Measures developed from the 10-K filings are considered to be more revealing in terms of firms' innovation strategies and activities as compared to their annual reports (Merkley, 2014). But, a decline in R&D-related sentences does not necessarily mean a decrease in the number of patents. Also, Huang et al. (2020) study a specific group of innovative entities, the publicly traded firms. 10-K is not applicable for individuals and small entities but these entities participate significantly in innovative activity.

Though compared to pre-AIA, Huang et al. (2020) find a decline in R&D-related sentences for publicly traded firms post-AIA, their market returns did not significantly change (Lerner et al., 2015). Among the venture capital-backed firms (VC), Lerner et al. (2015) do not find any significant difference in their formative stage funding post-AIA. It is unknown how the other small entities, such as individual inventors, small firms not backed by VCs, and universities filed their patents before and after the AIA. In fact, the quantity and quality of patents for all entities before and after the AIA remain

[14] Canada's reform: https://www.ic.gc.ca/eic/site/cipointernet-internetopic.nsf/eng/wr04732.html
[15] Phillipines' reform: https://wipolex.wipo.int/en/text/488675

understudied. Neither 10-K filings nor information on VC fund disbursement completely capture the patenting activity of entities, but they do provide a direction to this paper. The mixed evidence of a decline or insignificant change in the innovative activity while the AIA encourages disclosure seems counter-intuitive. However, economics suggests that given different conditions we can observe either an increase or a decrease in the disclosure of inventions. AIA rewards early filers, which may result in a direct increase in patent applications. But, if an entity discloses one of its inventions, its competitors can preempt any pipeline of inventions that they may have had, which may be connected to a focal invention. Because FITF rewards the first filer as opposed to the previous rule of FTI where an inventor could furnish proof of invention date if another inventor filed for a similar invention, an entity may become secretive and only file patent applications when they are sure that their patent application will not be invalidated or their future work cannot be preempted.

In the case of the AIA, Huang et al. (2020) discuss a similar mechanism of holding back on patenting because of competitors' preemptive activity and therefore we observe a decline in narrative R&D disclosure post-AIA. Abrams and Wagner (2013) also discuss similar explanations for the decline in patents by individuals post-Canadian reform. Some of the possible explanations they discuss are fewer resources for individual inventors which makes it difficult to file quickly for multiple patents, a shift in the use of different intellectual property protection mechanisms such as trade secrets, or a shift to patenting in the US. The authors find significant evidence of an increase in Canadian inventors patenting in the US after the Canadian reform. This is one of the reasons why inferences from the Canadian case cannot readily predict the effects of the AIA. In Canada, inventors still had an option of enjoying First-to-Invent if they filed their application in the US. Therefore, it is possible that the Canadian entities shift their activity toward the US, especially because of their geographical proximity, as well as the US, being a hub for innovative activities. However, in the case of AIA, entities do not have an option to move their activity to another country where they could still enjoy the FITF rule. Therefore, concluding that reforms similar to the AIA always reduce innovative activity, especially by small entities may be misleading.

The other strategies as described in Abrams and Wagner (2013) may play out in the US, especially for small entities. Rather than shifting the patenting activity to another country, entities in the US could choose higher secrecy and file patent applications only for those inventions which they expect to prevail in the market. Economic theory suggests mixed effects of the AIA on invention disclosure (Scotchmer and Green, 1990). Empirically, it is unknown if its intended effects of achieving certainty were indeed achieved or not (Lerner, 2000; Vandenburg, 2013; Cerro, 2014).

2.3 Conceptual framework

The various components of the AIA inevitably overlap with one another. Because of this, even if AIA's FITF rule targets a specific group of lawsuits, the other policy changes provide greater certainty in securing patent rights in the post-AIA period compared to the previous regime. Therefore, the AIA affects all patent applications and not only those which were prone to litigation on the invention date. It may be impossible to decompose the overall effect of the AIA into the effects of the individual components of the policy. But, we know that the AIA's primary goal which encompasses all its components is to improve the innovative environment in the US. The way chosen by the AIA to improve is by favoring certainty over flexibility. Greater certainty should result in timely disclosure of inventions, which also translates to an increase in the number of disclosures. The rationale behind this change in policy is that the increase in welfare from increased disclosure surpasses the welfare from a flexible date of invention. The broadest and most direct measure of innovation that stems from this reasoning is the count of patent applications. Since the AIA encourages disclosure of inventions, I expect an increase in the number of patent applications post-AIA as an indicator of increased disclosure of inventions. This increase is not only expected from the existing entities but should also encourage new entities to disclose their invention. This is the first question this paper explores, for small entities, did the number of patents increase after the AIA as compared to large entities and themselves before the AIA? A higher number of patent applications without an increase in their quality of would not mean much. In fact, the AIA specifically mentions that enhancing patent quality is one of its goals. However, the AIA's goal of encouraging quality disclosure can be met with the opposite effect because entities may respond strategically, and like every policy, the AIA is also laced with unintended consequences. It is possible that even though the AIA encourages invention disclosure, it increases the cost of disclosure of follow-on inventions. It is therefore unknown if the quality and quantity of patent applications indeed increase as a result of the AIA. This is the second question this paper explores, did the quality of patents by small entities increase after the AIA as compared to large entities and themselves before the AIA?

3 Data

3.1 Data sources

The main results of this paper are estimated using all the patents applied for at the USPTO between and including the years 2008 and 2016. Since not all entities release their balance sheet information, such as their R&D expenditure, assets, number of employees, etc., but such characteristics are determinants of patenting behavior, I use a subset of all patents that are applied for by the publicly traded firms in a separate analysis. For this subset of patent applications, I can control for certain entity-level characteristics (also called firm-level, since they are publicly traded firms). The data sources of patent-level and firm-level characteristics are given below:

Patents: Patent Examination Research Dataset (PatEx) is the dataset compiled by the Office of the Chief Economist (OCE) at USPTO that contains patent-level characteristics for the patents applied in the US from the 1900s to 2020. [16] In this dataset each row is a patent application which contains all the relevant information the patent application's prosecution generated till date; for example, the application number, filing date, issue date (if granted), number of claims, etc. Entity names in the patent database are

[16] Patent Examination Research Dataset: https://www.uspto.gov/ip-policy/economic-research/research-datasets/patent-examination-research-dataset-public-pair

not standardized and Patentsview bridges this gap using disambiguation algorithms to standardize patent assignee names. I sum the relevant statistics for each quarter for each assignee using the Patentsview standardized assignee names. [17]

Firms: Compustat's North America data provides quarterly financial information from the quarterly balance sheets of the publicly listed firms.[18] The Center for Research in Security Prices (CRSP) provides daily stock prices of these firms.[19] I average the daily stock prices to a quarterly-level.

While Patentsview standardizes patent assignee names, they still have to be connected to the Compustat database to get their financial information. Also, firms merge or hold subsidiaries that may individually file for patents. They may be listed as a different firm in the patent database but the patent belongs to the parent firm. Kogan et al. (2017) (henceforth referred to as KPSS) and Arora et al. (2021) (henceforth referred to as ABS) bridge this gap by standardizing the firm names, connecting them to the patent database, and connecting the Compustat firms and their subsidiaries to one standardized name. I use their databases to obtain a sub-sample of publicly listed firms' patents.

Litigation: The Office of the Chief Economist (OCE) at USPTO recently compiled a dataset on cases involving patents filed at district courts in the United States from 1963 to 2016. This dataset is called the Patent Litigation Docket Reports (PLDR). From the PLDR, I create a measure of exposure to litigation at the entity-quarter level.

3.2 Sample selection

There are three types of patents in the US: utility, design, and plant. Inventions relating to new products or processes, or their improvement are utility patents. The AIA's objectives are best represented by these inventions and therefore, in this paper, I only

[17]Patentsview is a collaborative project developed by the USPTO, American Institutes for Research (AIR), University of Massachusetts Amherst, New York University, University of California, Berkeley, Twin Arch Technologies, and Periscopic. See: https://patentsview.org/what-is-patentsview

[18]Compustat is accessed through Wharton Research Data Services (WRDS). https://wrds-www.wharton.upenn.edu/pages/about/data-vendors/sp-global-market-intelligence/

[19]Center for Research in Security Prices (CRSP) is accessed through Wharton Research Data Services (WRDS): https://wrds-www.wharton.upenn.edu/pages/about/data-vendors/center-for-research-in-security-prices-crsp/

study the utility patents. Plant or design patents are granted to inventions that relate to the development of new plants and new designs, which may be unaffected by the aspects of the AIA this paper focuses. [20] Each row of the PatEx database is a patent application. The 2020 release of PatEx contains 16,514,638 patent application numbers. After removing design, plant, and blank patent applications, I have 15,811,897 patent applications. A patent application can appear multiple times in the dataset through continuations but will culminate in *one* granted patent.[21] A patent application appearing for the first time in the patent database is called a "parent" application, and all the connected applications, appearing later, are called its "children" applications. Multiple parent applications can be connected to multiple children applications. Using the parent and children continuation data from PatEx, I connect all the parents and their children to find the earliest application date. 13,537,926 patent applications contain a filing date.[22] I also restrict the dataset to patent applications that were applied between and including the years 2008 and 2016. I take patent applications post-2008 to avoid distortions from the financial crisis and till 2016 for comprehensive coverage of patent applications. The average grant lag of a patent is about four years and two standard deviations above the grant lag are about eight and half years at the USPTO.[23] Therefore, as years go by, the

[20] See USPTO patent process for details on the definitions of utility, plant, and design patents here: https://www.uspto.gov/patents/basics/patent-process-overview#step3

[21] At the USPTO, a patent application can be initially filed as a provisional or a non-provisional application. A provisional application may not contain claims or the specifics of the invention. Its primary use is to establish an effective filing date and should be followed by a non-provisional application, applied within 12 months of the provisional application's filing date. A non-provisional application is prosecuted by an examiner to determine its patentability. This type of application can further be continued as a continuation application, in-part continuation application, or divisional application can If an entity wishes to apply to more than one country, they may opt for a Patent Cooperation Treaty (PCT) application. If applied as a PCT application, the applicant has to choose the countries they wish to apply for patent rights. If an applicant wishes to file for patent protection in multiple countries, rather than applying to every country separately, the entity may choose to file the application as an international patent. This type of patent is also called a Patent Cooperation Treaty (PCT) application. The decision on grants is still given by the countries separately. For details refer here: https://www.wipo.int/pct/en/faqs/faqs.html Later, this enters the conventional application procedure, during which a new application number is assigned to the provisional or PCT application. Counting a PCT or a provisional and its conventional application counterpart as two different patent applications will result in double counting.

[22] The patent application that does not contain a filing date usually are filed as a PCT application. In place of the filing date, such applications receive a World Intellectual Property Organization (WIPO) publication date. The USPTO records these patent applications as National Stage Entry (NST) when they are examined at the USPTO, they receive an application number, and they do contain a filing date at this point.

[23] Grant lag winsorized at the 5 and 95 percentile cutoffs post-2000.

proportion of pending patent applications post-2016 would increase. For pending patent applications, information on citation is unavailable or scarce. The verdict for these applications, whether they will be granted or abandoned, is also unknown. To avoid these, I consider the patent applications that are applied on or before the end of 2016.

Restricting patent applications by year leaves me with 5,087,133 patent applications. Among these, 3,255,080 patent applications have a standardized assignee ID or an inventor ID in the Patentsview Database. The rest, 1,832,053 patent applications, do not have any ID. While these patent application numbers are unique in terms of their numbers or labels, they still can be a derivative of another application, as discussed earlier. In Appendix A.1, I provide a detailed discussion of the patent application numbers that do not have any ID and provide reasons as to why these applications are either insignificant or are repeats of the patent applications already in the main sample of patent applications that can be identified with a standardized ID.

There are 581,980 unique entity IDs that have applied for patent applications between and including the years 2008 and 2016. I use the standardized Patentsview assignee IDs, wherever they are available, and the standardized inventor IDs where the assignee IDs are unavailable. For the rest of the paper, I refer to the entity identifiers as entity IDs. An entity can be identified as either a company, or an individual, or a Government entity. Assignee IDs are unavailable when an applicant is an individual. Each ID on average are involved in 5.6 patent applications, which includes joint work. Using the quarter of the earliest application date of each patent and the entity IDs, I sum or average all the variables at the entity-quarter level. I detail all the variables in the next section. 3,259,984 patents by 581,980 IDs are reduced to an unbalanced panel of 1,852,143 entity-quarter rows. Among these, 703,692 entity-quarters appeared only once in the sample, and therefore cannot be compared within entity. These observations are singletons. Singletons in models with entity fixed effects where the standard errors are clustered may overstate the number of clusters and hence the statistical significance (Correia, 2015). I, therefore, report the main results without the singletons. The results of this paper however remain unchanged even with the inclusion of the singletons, as we will observe while discussing the results in Section 5. The final sample for the main set of analyses, therefore, is 1,148,451 entity-quarter observations. The concern is evident about entities self-selecting themselves into pre and post-AIA depending on their objectives. The unbalanced panel used in estimating the main results does not have zeroes when an entity does not file for any patent application for a given quarter. I address this issue by adding zeroes for the quarters where an entity does not patent, thereby balancing the entity-quarter panel and replicating the main analyses in Appendix A.2.

Next, I use patent number-*PERMNO* match from Kogan et al. (2017) to identify patents by publicly traded firms.[24] *PERMNO* is the permanent issue identifier as provided by CRSP data. There are 2,075 unique *PERMNO*s in Kogan et al. (2017) who had applied for patent applications between 2008 and 2016. I also use Arora et al. (2021) to match the *PERMNO*s with the Compustat identifier, *GVKEY*. There are 1,374 unique *GVKEY*s who had applied for patent applications between 2008 and 2016. For the firm-level analyses, I have 1,374 publicly traded firms' characteristics and their patent applications. Different studies employ their own assignee standardization method but Arora et al. (2021) corrects for mergers, acquisitions, and patent reassignment. From the 3,255,080 patent applications, 1,309,860 patents can be identified as having a *PERMNO* and a *GVKEY*. These observations constitute the sample for a separate set of analyses in Table 8.

3.3 Variables

3.3.1 Outcomes

I use two patent outcome measures as dependent variables to study the disclosure and the value of invention; log(number of patents) and scaled citations within two years of issue of a patent.[25] Older patents may have a higher number of citations and are incomparable to the newer patents; and number of citations may vary by technology classes and years (Mezzanotti, 2021; Lerner and Seru, 2022). Because of these two reasons, they are capped

[24]The authors have released data updated till 2020. See: https://github.com/KPSS2017

[25]If a patent application is applied by N assignees, then each assignee receives $\frac{1}{N}^{th}$ of the patent.

within two years of issue and adjusted by United States Patent Classification (USPC) subclass and year.[26] These values are aggregated at the assignee-quarter level.

Literature on innovation had routinely transformed number of patents into log(1+number of patents). But recently, Chen and Roth (2022), and Mullahy and Norton (2022) have questioned the addition of 1, which changes the shape of the underlying distribution. To tackle this, I only log-transform number of patents, without the 1. Since there are no zeroes, log of number of patents provides a monotonic transformation and allows interpretation of coefficients in percentage changes. Measurement of citations are slightly more complicated than the number of patents. Patents may as well receive zero citations, which cannot be log-transformed. Additionally, innovating entities have over the course of time changed their strategy of citing other patents, and the overall citation have increased in recent years. This also varies by the various fields of inventions. These complexities have been documented by Lerner and Seru (2022), Higham et al. (2021), and Jaffe and De Rassenfosse (2019). Therefore, I calculate the percentile of citation within an NBER subcategory and year.

Using the parent application number's date of filing, I create an indicator called *Post*. This variable takes a value of 1 if the patent application was applied for after March 16, 2013, and 0 otherwise. This is the date when the AIA's FITF was enacted. But, as describe in Section 2, The AIA's was signed in Sept 16, 2011. Some of the other provisions were enacted at this time, while the main legislation, the adoption of FITF rule, was implemented in March, 2013. I use this date as the primary cut-off to define the pre and post periods, but in the later part of the paper, and as a part of robustness, I report results using both the the AIA's signing date, as well as the AIA's FITF implementation date.

Note that to determine if a patent was applied for in the pre-AIA or post-AIA period, I use the "parent" application number as opposed to the "patent" application number. This is because a patent application can claim priority to an earlier provisional or non-provisional application. Through this, an application filed later claims continuation from

[26] Average scaled citations are defined as the average of percentiles of citations by NBER subcat and year.

a previously filed application. If this happens, then the earliest application date is considered the application date for the patent application. In such cases, the patent applications filed post-AIA can still be under the purview of pre-AIA rules.

PatEx provides an indicator *Small Entity*, which takes a value of 1 if the patent application was filed by a small entity and 0 otherwise. A small entity is defined as a person, a small business concern, or a nonprofit organization, which includes universities and educational institutes. An entity is considered to be a small business concern if it meets the size standards, i.e. the total number of employees and affiliates should not exceed 500 persons at the time of application. The size is self-declared at the time of patent application and is not formally verified by the USPTO. A small entity receives a 50 percent discount on all patenting fees. An entity can file its patent as a small or a large entity, and the reasons for choosing either are unknown. [27] To keep the main results simple, I assume that an entity is small if 50 percent or more number of patents out of all patents by the entity is filed as a small entity. In Appendix A.4, I relax this assumption to redefine small entity using the proportion of patents filed as small among all patents an entity had filed for that quarter and replicate the main results.

3.3.2 Controls

The two most important controls in this paper are the entity and quarter fixed-effects. While quarter fixed effects are routine, entity fixed effects were difficult to control for due to variation in the spelling of assignees. I explain the use of Patentsview IDs in Section 3 to mitigate this problem. Entity fixed effects allow us to estimate the average change at the entity level.

A patent application is categorized in the United States Patent Classification (USPC) System. Each patent application is assigned a USPC class and subclass. Since there are numerous USPC classes, I aggregate them into six NBER categories and 36 NBER subcategories. The broad six categories are chemical, computer and communications, drugs

[27] If an entity that qualifies as a small entity transfers the rights of its patent to an undiscounted entity partly or fully, then the patent will cease to be from a small entity. But, licensing to a federal agency or using the patent as a security interest does not preclude the entity's patent to be considered as one from a small entity.

and medical, electrical and electronic, mechanical, and others.[28] Since I am collapsing the patent-level statistics into the assignee-quarter level and an assignee can have patents in various categories in each quarter, I take the mode category. If there are multiple modes of categories, I randomly choose one category for that assignee-quarter.

Among the patent level controls, literature using patent-level statistics use the number of claims as a measure of complexity and scope of the patent. Claims are specific statements that define the uses of a patent. In general, the higher the number of claims, the greater the patent's number of defined uses. Therefore, it becomes progressively difficult to add more and more claims to one patent application, as it adds multiple uses in one patent. Until recently, claims were thought to be a mix of complexity and quality of patents, but Marco and Miller (2019) dispel a few confusions around the significance of claims. They find that the narrow claims have a greater probability of being granted. Along with this, claims' length and breadth are strategically written and fought with the examiner. Since the AIA changes incentives to file for patent applications substantially, all the variables which capture patenting strategy are subject to change. This paper's focus is limited to studying the two broadest measures of quantity and quality of inventions, and therefore, I do not comment much on the claims of patents, and their constituents, as they deem a separate study. But in Appendix A.9, I report the gap between small and large entities' log(avg. number of claims) post-AIA. We observe a significant drop in the mean claims post-AIA by small entities. A drop in mean claims suggests that each patent post-AIA has on average lower number of defined uses. This also suggests that patent applications are now shorter in length. This is why claims are not a good covariate that should be in the list of control variables, as it would absorb significant variation in the number of patents and citations.

I use the log experience of an applicant at USPTO. I also control for the number of patents that are maintained at the 4^{th} year post-grant, or abandoned. Maintenance or renewal of patents and abandonment are predictors of the quality of patents. The higher the quality, the greater the chance that the patent would be renewed by the assignee,

[28] A concordance between NBER categories and USPC classes are provided in Hall et al. (2005)

and on the contrary, the lower the quality the higher the chance that the patent would be abandoned by the assignee (Bessen, 2008). Next, following De Rassenfosse and Raiteri (2022) and Webster et al. (2014), I develop a measure for expertise in a field for an entity, called Revealed Technological Advantage (RTA). RTA is defined as the share of patents by an applicant over the share of patents by all applicants in a patent technology class. If an applicant applied the bulk of their patent applications in a given technology class that is higher than the share for all entities in the same technology class, the ratio will be greater than 1. Therefore, an RTA value greater than 1 signifies high expertise of an entity in the given class, while the opposite, an RTA value lower than 1 signifies low expertise. These controls act to provide a stricter restriction to the estimates.

From Kogan et al. (2017) and Arora et al. (2021), I obtain a patent-*PERMNO-GVKEY* match. I use this information to create a subset of patents summed at the firm-quarter level. For these firms, I observe different firm characteristics which come from their quarterly balance sheets. Following the literature, I control for their quarterly log number of employees, log number of assets, log R&D expenditure, and log of firm age since establishment (Hegde and Sampat, 2009).

3.3.3 Litigation exposure

Following Mezzanotti (2021), I define exposure to litigation as a weighted average of litigation in all NBER subcategories, where the weights are the shares of patents by an entity in a subcategory at a given time.

I calculate the proportion of litigation in each subcategory till time T, where $T = 1 + 2 + ... + t$ as the number of litigated patents in subcategory c till time T (L_{ct}) over the number of patents in subcategory c till time T (P_{ct}). Formally:

$$l_{cT} = \frac{L_{cT}}{P_{cT}}$$

Next, I calculate the proportion of patents in each subcategory c by entity i at time t as the number of patents by entity i in subcategory c at time t over the number of patents by entity i at time t. Formally:

$$p_{ict} = \frac{P_{ict}}{P_{it}}$$

To obtain entity level measure of exposure to litigation (e_{ict}), I calculate the average litigation till time T for all subcategories i.e. average of l_{cT} for all the 36 NBER subcategories and weigh each of the subcategories with the proportion of patents by an entity i in those subcategories at time t (p_{ict}).[29] Formally:

$$e_{ict} = \sum_{c=1}^{36} p_{ict} \times l_{cT}$$

[29]The NBER categories and subcategories are explained in Subsection 3.3

Table 1: Definition of the variables

Panel A: Quarterly patent data		
Variable	Definition	Source
Log (no. patents)	Log(number of patents)	PatEx
Scaled 2 yr citations	Mean of citations' percentile by USPC subclass and year where citations are within 2 years of issue	Patentsview
Post	1 if the patent was filed on or after March 16, 2013	
Small entity (SE)	1 if the patent was filed by a small entity; 0 otherwise	
Log experience	Log experience of assignee at the USPTO	
First patent	1 if the patent was the first patent for the assignee; 0 otherwise	PatEx
NBER category	Patent classified into one of the six NBER categories: chemical, computer and communications, drugs and medical, electrical and electronic, mechanical, and others	
Renewal proportion	Proportion of patents renewed (maintained) at 4^{th} year	
Pending proportion	Proportion of patents pending as of Dec 2020	
Abandoned proportion	Proportion of patents abandoned	
Joint patent	Proportion of patents jointly applied	
Entity type	0 if company, 1 if individual, and 2 if Government	
Revealed tech. adv.	Revealed technological advantage is a ratio of ratios the proportion of patents in category c (P_{ict}) out of all patents by entity i at time t (P_{it}) over the proportion of patents in category c (P_{ct}) out of all patents by all entities at time t (P_t) Formally: $\frac{\frac{\sum P_{ict}}{\sum P_{it}}}{\frac{\sum P_{ct}}{\sum P_t}}$	PatEx
Litigation exposure	Weighted average of shares of litigation litigation till time T in category c (L_{cT}) out of all patents till time T in category c (P_{cT}) weighted by patents by entity i in category c at time t (P_{ict}) out of all patents by entity i at time t (P_{it}) Formally: $\sum \frac{P_{ict}}{P_{it}} \times \frac{L_{cT}}{P_{cT}}$ where the time till date (T) is defined as $T = 1 + 2 + ... + t$	PLDR

Panel B: Quarterly firm data		
Variable	Definition	Source
Log emp	Log number of employees	
(Log assets)/emp	Log assets per employee	Compustat
(Log R&D)/emp	Log R&D expenditure per employee	
Log age	Log age since establishment	
Log Market cap	Log(stock price × shares outstanding)	CRSP
KPSS value	Log real-KPSS value of patents	KPSS
PERMNO	CRSP firm identifier; identifies firms in KPSS	
GVKEY	Compustat firm identifier, identifies firms in Compustat; ABS provides GVKEY-PERMNO match	ABS

Panel A lists the variables used for the assignee level analyses. Panel B lists the variables used for the firm level analyses. Kogan et al. (2017) is abbreviated as KPSS, and Arora et al. (2021) is abbreviated as ABS. The variables *PERMNO* and *GVKEY* are not variables. From the assignee-quarter level patent data, the firm identifiers PERMNO and GVKEY are used to create a sub-sample of only-firm assignees.

3.4 Descriptive statistics by small and large entities

In this section, I describe a few crucial differences between small and large entities from the sample constructed for this paper. This helps us understand small entities in the context of this paper contrasting it with the literature that have previously studied small entities in certain capacity. I additionally provide a summary of these variables in Table 2.

Table 2: Differences between small and large entities

Variable	Small	Large
Patents	1.67	4.70
Citations	0.48	2.06
Abandoned	0.53	0.25
Cases	0.05	0.03
Exposure	-0.014	0.004
RTA	3.69	1.81

Patents and citations are average patents filed and citations received within two years of grant by small and large entities. Abandoned and cases are proportion of abandoned patents and proportion of patents involved in district court cases for small and large entities. Exposure is a standardized measure of exposure to litigation and RTA is a measure of expertise. The means exposure and RTA are reported in this table. See Subsection 3.3.3 and Table 1 for detailed definitions. All the values are calculated using the paper's sample.

A typical small entity on average files for 1.67 patents in their lifetime, while a typical large entity on average files for 4.7 patents. While the average patents may seem low, the top patenting entities for each of these categories can file upto 4500 and 50,000 patents respectively, which illustrates the long tail of the distributions for patenting entities. The long tail is similar for both small and large entities. On similar lines, a typical small entity receives 0.48 citations while a typical large entity receives 2.06 citations within the first two years of their patents' grant. These numbers are consistent with prior literature, showing that small entities receive lower citations on average compared to large entities. However, an interesting pattern lies in the tails of the distribution of citations for small and large entities. The 95^{th} percentile cutoff of citations received by small and large entities are — 1 and 0.5 respectively. This indicates that 95 percent of small entities receive higher citations than 95 percent of large entities. The heavy-citation-receiving large entities pull the average for large entities so much that they on average surpass small entities. In other words, most small entities patent inventions that can be considered of higher quality than most large entities.

Along with filing for lower number of patents and receiving lower citations than large entities on average, small entities also do not renew, or – abandon 53 percent of their patents, which for large entities are 25 percent. This is also consistent with the literature, and the reasons cited are often times financial in nature (Bessen (2008)). Next, patents by small entities are also involved in a greater number of district court cases — about 5 percent of patents by small entities in this paper's sample were involved in cases between 2000 and 2016, while the same is 3 percent for large entities, as also discussed in Marco and Miller (2019) and Lanjouw and Schankerman (2001). Interestingly, if we look at average standardized exposure, we see that large entities file for patents more in the fields that have a higher exposure to being litigated. Finally, small entities are also highly specialized in their field as compared to larger entities, as shown by the average of Revealed Technological Advantage (RTA). A greater RTA means the proportion of patents by a particular entity in a particular field out of their portfolio of patents is greater than the proportion of patents by all entities in a particular field — also indicating that small entities put most of their eggs in one basket.

3.5 Descriptive evidence around the AIA

Figure 1 shows the number of applications filed each month at the USPTO since 1975. It also marks the months when the said amendments to the Patent Act were enforced. We observe spikes in patent applications during or after each amendment. After each spike, we also observe either a change in growth rate or a parallel shift in monthly patents or both. Along with the amendments to the Patent Act, certain patent lawsuits also delineate boundaries of patent rules by establishing case laws, which may affect patenting activity but such cases are not marked in the figure. The AIA came into effect as a law on September 16, 2011, which is the second red dotted line from the right. The FITF rule came into effect on March 16, 2013, shown by the rightmost red dotted line. A large spike in patenting can be observed in the month when the AIA's FITF rule was enforced. From the Great Recession to AIA, the growth in the number of patents was positive which dampened after the AIA spike. But, we can observe an increase in noise post-AIA. I investigate this further in Figure 2.

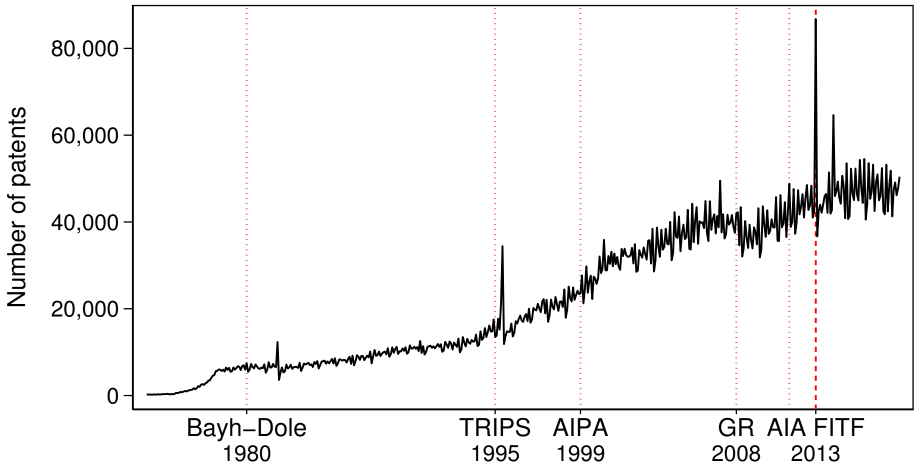


Figure 1: Patent applications filed each month
This figure plots total patents for each month from 1975 through 2018. The months of enforcement for the policies are Bayh-Dole Act (Dec 1980), Trade Related Intellectual Property Rights (Jan 1995), American Inventors Protection Act (AIPA; Nov 1999), Great Recession (Sept 2008), and America Invents Act (Sept 2011 and Mar 2013). Coverage of patents before 1981 is poor and only granted patents are observed till 1999. Post-1999, due to the introduction of pre-grant publication of patent applications, we see all barring the classified patents.

Figure 2 furthers the investigation of the patenting activity around the AIA by separating the patents into small and large entities. After a preliminary cleaning of raw patent application statistics, such as removing plant and design patents and restricting the dataset to observe patent applications only applied between 2008 and 2016, I plot Figure 2. It zooms in on the portion of Figure 1 after the Great Recession and plots the change in quarterly patents for small and large entities. I move from the monthly number of patent applications to quarterly because of two reasons. First, it irons out large monthly variations, making it easier to observe the pattern, and, second, all the empirical analyses are done at the entity-quarter leve. Also, for publicly traded firms only quarterly financial statistics are available. In the estimations I also switch from the absolute number of patents to log(number of patents) because regressing large absolute values on small $X's$ or vice-versa makes the estimates unreadable, and we are interested in studying the percentage change. But in the descriptive evidence figures, I plot changes quarter-on-quarter for the absolute number of patents.

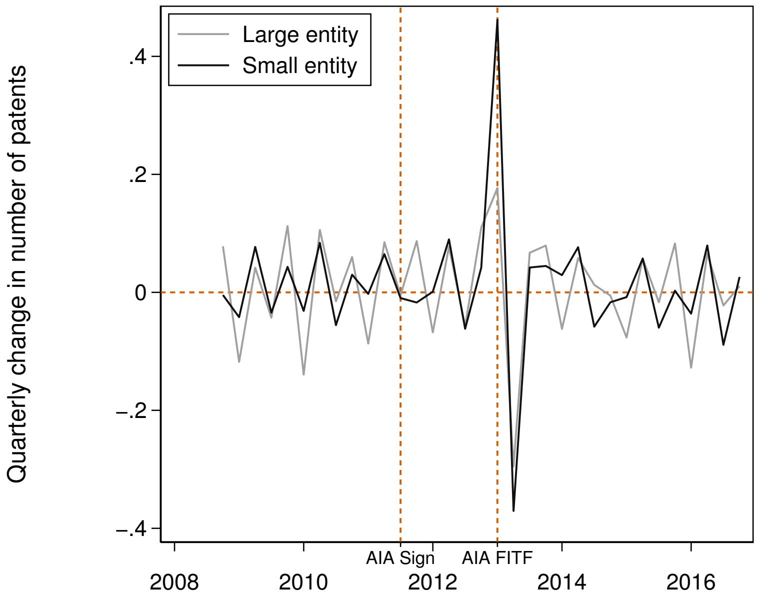


Figure 2: Patent applications by small and large entities
This figure plots quarter-on-quarter change in the absolute number of patents for small and large entities from 2008 to 2016.

We can observe similar trends for both small and large entities in Figure 2. In the USPTO database, entities are recorded as small entities or undiscounted entities. Undiscounted entities do not receive discounts in patenting fees and therefore are assumed to proxy large entities. The number of patents filed by large entities is always higher than the number of small entities, but the change quarter-on-quarter remains similar. Similar to Figure 1, quarterly patent applications grow quarter-on-quarter till the AIA's enforcement, and then the growth rate nears zero.

This fails to highlight the following: the average change pre and post AIA within entities, and if entities who predominantly file for high number of patents on average are different from the low patentees. Number of patents is a positively skewed distribution with a substantially long tail. A large number of small inventors and firms reside at the start of the distribution, having about one or two patents each quarter, and as we go along the x-axis of that distribution, we find large firms with around 2000 patents each quarter. Therefore, I show the main results of this paper considering only subsamples of entities by removing top and bottom percentile of the distribution where they operate.

Not all patent applications are drafted equally. Some are of greater quality than others. Figure 3 plots the change in the number of citations per patent in the first two years after the issue of a patent quarter-on-quarter. Older patents will have an advantage here, and to limit that, I restrict the citations for the first two years after a patent's issue. Note that in the main set of analyses, I use a different measure of citations, called average scaled citations, as explained in Subsection 3.3. The quarter in the x-axis is the patent application quarter, rather than the issue quarter since a patent issued post-AIA can still be applied for before the AIA. Small entities on average have a lower number of citations, similar to a lower number of patents. One of the key differences we observe here is that the change in citations start responding immediately after the AIA's sign, while the number of patents change only at the FITF's implementation. Variation in citation among small and large entities is also more than the variation in the number of patents.

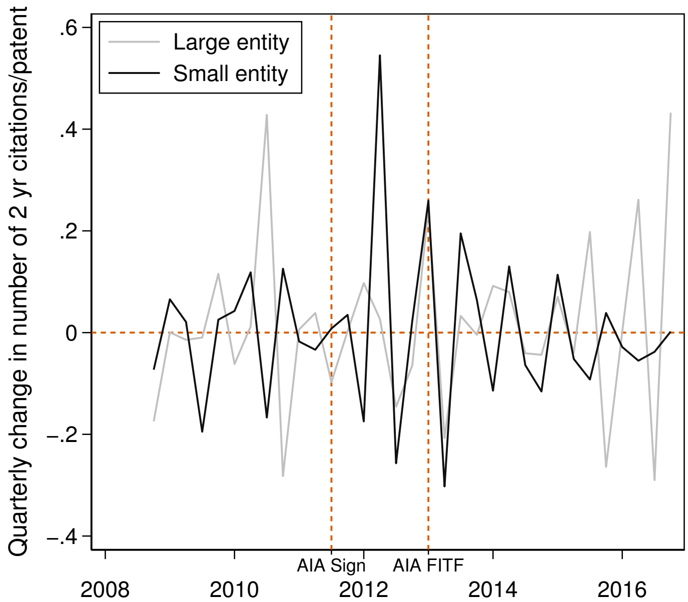


Figure 3: Citations/patent within two years

This figure plots quarter-on-quarter change in the absolute number of citations per patent within two years of issue for small and large entities from 2008 to 2016.

4 Empirical strategy

I estimate the change in the log(number of patent applications) and average scaled citations at the assignee-quarter level using a difference-in-differences method. The number of patent applications and the adjusted percentile of citations proxy quantity and quality of inventions respectively. I compare the change in quantity and quality of inventions by small entities post-AIA with the inventions by small and large entities pre-AIA.

The main set of results, as presented in Tables 3 and 8 use two different samples: all patenting entities and a subset of them — the publicly traded firms involved in patenting, respectively. Next, I develop a measure of exposure to litigation, following Mezzanotti (2021) and estimate the changes in the two outcomes for only small entities. I use the measure of exposure to litigation as a continuous treatment, and the results are reported in Figures 7a and 7b.

A crucial assumption of this paper hinges on is the comparison of small and large entities pre and post-AIA. I assume that if the AIA was not enacted, there would be no change in the rate of change of the number of patents and citations between small and large entities quarter-on-quarter. Also, by comparing discrete entity sizes i.e. small and large entities, I implicitly assume that size is an adequate measure to capture the resources the entities have in their disposal. This measure because of being discrete dampens the variation that I could have exploited in the estimation if I had a continuous measure of resources available to entities.

The main set of results i.e. the entity level and the firm level analyses, are estimated using Equation 1, and the full specification is as follows:

$$Y_{it} = \beta_0 + \delta(Post \times SE_i) + X_{it}^{'}\beta + \lambda_i + \lambda_t + \varepsilon_{it} \tag{1}$$

Here Y_{it} denotes two outcome variables, the log(number of patents) for an assignee at each quarter and the average scaled citations received by an assignee at each quarter within two years of the patents' issue. The coefficient δ captures the change in the difference in Ys between small and large entities after the implementation of the AIA. I control for the assignee i and the quarter t's baseline using assignee and quarter fixed effects, given by a range of indicator variables and their coefficients in the matrices λ_i and λ_t respectively. I also control for a range of patent quality and complexity correlates denoted by the vector X_{it}. A discussion on these measures follows after the models used in this paper are explained.

I estimate this equation for the full sample, i.e. all the patenting entities, and for only the publicly listed firms. For the firm-level analyses, I employ additional firm-level controls, which control the firm's resources with greater precision than only using the information derived from their patenting behavior and entity size.

While δ reports the average of Ys for small entities over all the quarters post-AIA, I also separate the effects by each quarter. This enables us to understand and verify if any pre-trend influences δ. The estimating equation is given by Equation 2. The full specification is:

$$Y_{it} = \beta_0 + \sum_{s \neq 0} (\beta_s \times 1[s = t] \times SE_i) + X'_{it}\beta + \lambda_i + \lambda_t + \varepsilon_{it} \tag{2}$$

In this Equation, β_s ranges from the first quarter of 2008 to the last quarter of 2016, barring the first quarter of 2013 i.e. when the AIA was implemented, which acts as the base quarter. The other coefficients have the same interpretation as in Equation 1.

The crucial assumption, as described previously, is the use of large entities as a comparison group for small entities. Therefore, I try to report the results from different perspectives, and argue that given the controls variables, large entities can act as a comparison group for small ones.

Next, following Mezzanotti (2021), I estimate the change in the log(patents) and avg. scaled citation for varying degrees of exposure post-AIA. The full specification is given by:

$$Y_{it} = \delta(Exposure \times Post) + X'_{it}\beta + \lambda_i + \lambda_t + \varepsilon_{it} \tag{3}$$

Exposure is defined in Section 5.3, and is standardized. Therefore, δ reports the average change in the log(number of patents) and avg. scaled citations post-AIA when exposure to litigation increases by one standard deviation. Note that in this equation, I do not distinguish between small and large entities. Though I do estimate this for only small entities, and for all entities. Later, in a triple-difference setup, I show how exposed small entities post-AIA compare with the unexposed small and large entities, and report the results in Appendix A.11.

5 Results

5.1 Evidence from all small entities

In this section, I examine and report the change in two outcome variables: quantity of inventions — log(number of patent applications) and the quality of inventions — average scaled citations after the enactment of the AIA for all small entities in the sample. The results are reported in Table 3, and this table constitutes the main and broadest result of this paper. These results are also disaggregated by quarters and reported in event study forms in Table 4, Figures 4 and 5. In this section, I discuss those broad results, i.e. how the quantity and quality of inventions changed for small entities before and after the AIA — on average and quarterly; and compare its similarities and dissimilarities with the previous studies that focus on the AIA and AIA-like reforms in other countries. In further sections, I delve deeper into the nuances of these results. For this and all the subsequent tables, I present four columns for each Y variable, where the first column is with and the second is without controls.

The estimate δ from Equation 1 reports how the gap between small and large entities changes after the AIA's enactment. In Table 3, this is reported in the first row $SE \times Post$.

Here post is defined as the first-inventor-to-file's adoption date, i.e. March 2013, and the timeline for which the estimates are calculated are post-Great Recession i.e. 2008, to 2016. The difference between the number of patents filed by large and small entities is around 4.4 percent after the AIA's implementation. This indicates a relative drop in the number of patents filed by an average small entity as compared to an average large entity post-AIA.

Table 3: Main results: log(no. of patents) and avg. scaled citations

	Patents		Citations	
	(1)	(2)	(3)	(4)
SE x Post	-0.0557***	-0.0412***	-0.1689***	-0.2480***
	(0.0042)	(0.0030)	(0.0568)	(0.0562)
Controls	No	Yes	No	Yes
Qtr F.E.	Yes	Yes	Yes	Yes
ID F.E.	Yes	Yes	Yes	Yes
Subcat F.E.	Yes	Yes	Yes	Yes
N	1148250	1146957	1148250	1146957

This table reports the gap in the patents and citations between small and large entities post-AIA. The number of patents are measured by log(number of patents) and citations are measured by the average scaled citations within 2 year of issue of patent. The estimates are derived from the model $Y_{it} = \delta(Post \times SE_i) + X_{it}^{'}\beta + \lambda_i + \lambda_t + \varepsilon_{it}$, where i and t denote entity and quarter respectively, X_{it} denotes a range of patent and entity level controls as defined in Table 1, and λ_i, λ_t control for entity and quarter fixed-effects respectively. The entity-quarter level dataset consists of all patents applied between 2008 and 2016. The indicator variable SE takes value 1 if more than 50 percent of the patents for an entity was applied as a small entity, and 0 otherwise. *Post* takes value 1 if the patents were applied on or after the first quarter of 2013 i.e. the implementation of first-inventor-to-file rule in the United States.

Columns (1) and (3) report results without and (2) and (4) report with controls. P-values at 1, 5, and 10 percent are denoted by *, **, and ***. Standard errors are clustered at entity-level.

When we look at the differences in log(number of patents) for each quarter compared to the base quarter i.e. the first quarter of 2013 in Figure 4, a visual evaluation reconfirms the results from Table 3 — a drop in the number of patents by small entities post-AIA. While on average the gap between small and large entities increase, the quarterly estimates also fluctuate substantially. In particular, compared to Q1 2013, number of patents did go up until 2010, and post that, declined on every quarter, going in the negative territories post-AIA's first-inventor-to file rule's adoption. It should be reiterated that in 2011, the AIA was formally enacted. A narrower window of analysis keeping 2011's 3^{rd} quarter as the base quarter will be discussed in detail in Section 5.2.

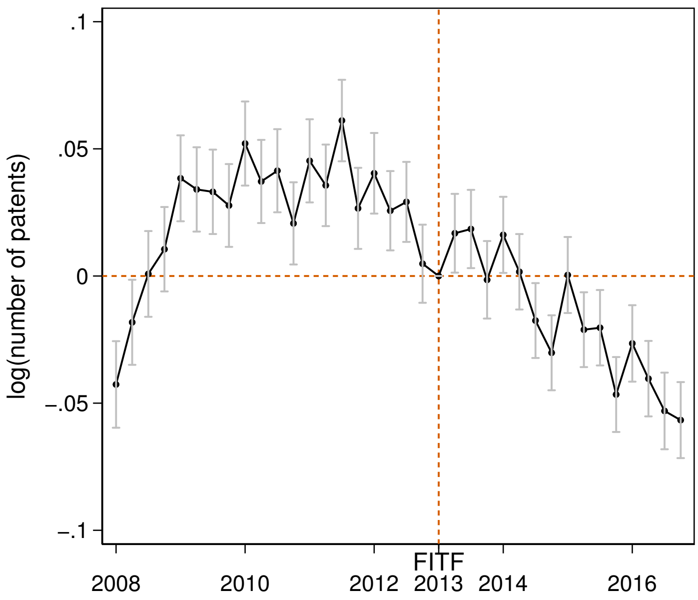


Figure 4: Change in small entities' patents over time
This figure plots the change in log(number of patents) by small entities for every quarter in our study timeline. The event study specification is reported in Equation 2, and the figure plots the β_s's estimated from $\sum_{s\neq 0} \beta_s \times 1[s = t] \times SE_i$ part of the equation, with the base quarter being the first quarter of 2013, when the AIA's FITF came in force.

This drop in patenting by small entities is similar to the one observed in Canada, when a similar change in patenting rule was enacted, i.e. a first-inventor-to-file rule. Abrams and Wagner (2013) and Lerner et al. (2015) note a drop in patenting activity for all types of inventors following enactment of such a legislation, and also report a greater drop especially among the individual inventors of Canada. The reasons for this, they note, are: one, firms have advantage in the "race" to reach the patent office first in terms of resources compared to the individual inventors, and on the contrary, these inventors not having the adequate resources in terms of fees to attorneys and agents that is required to file for patents quickly. Two, less invention by inventors and a shift to secrecy in order to protect their invention, which can also be considered as a corollary to the first reason, i.e. because of less resources to file for patents quickly after invention, small inventors have moved to a greater degree of secrecy than before. Three, Abrams and Wagner (2013) note that small inventors may be demoralized in filing for patents.

I do find certain evidence from the results from exposure to litigation, which pushes us to think in the direction of not only "demoralization" but also "testing the waters" after such a sweeping change brought forth by the AIA. This pathway will be discussed in detail in the Subsection 3.3.3. The next two reasons; four and five are: they speculate that individual inventors join firms, and shift their patenting activity to the U.S. This paper's scope and resources does not allow to test the former hypothesis of individual inventors joining firms, and the second hypothesis does not apply to the study of AIA, because the United States was the last country to adopt FITF rule. That means inventors neither have any other country where they can enjoy a rule of first-to-invent, neither do they have the geographical proximity such as moving their activity from Canada to the United States.

Before we move to discuss the nuances in the drop in number of patents by U.S. small entities, and their quality in the subsequent sections, let me comment on the other coefficients of the first two columns. The time trends are controlled using quarter fixed-effects. To minimize clutter, I plot the fixed-effects in a separate graph and is shown in Appendix A.5's Figure A8a and Figure A8b. The trends exhibit a pattern similar to Figures 1 and 2, i.e. for the quarters after the AIA, we observe a decline in the growth rate quarter-on-quarter and a plateauing of the number of patents for all the entities. Note that in Equation 1, SE and ID fixed-effects capture the overall effects for small entities and each entity's mean number of patents. The quarter fixed-effects report a trend over and above the individual baselines of each entity as compared to the quarter when the AIA was enacted.

The AIA is likely to induce a strategic change, as we observe with the number of patents. This may as well be observed in other measures that are derivatives and details within a patent document, and inclusion of those variables as a covariate in the model can absorb some of the variation in Y and therefore are "bad controls". Such a control, which is also an important measure of complexity is "claims" that is often used in the literature on the economics of patents. As discussed in Section 3.3.2, use of claims as a strategy is documented by Marco and Miller (2019). A change in incentives to patent due to the AIA's introduction can not only drop the number of patents as a response, but each of those patents could also be shorter and of a different quality. And, this is what is observed when I estimate change in log(claims) using the same specification as for the log(number of patents), i.e. Equation 2. This result is reported in Appendix A12. We can see from the figure that the number of claims per patent do fall after the AIA's enactment. Since it is a per patent measure, and since we know that the number of patents as well drop after AIA, it must be the case that the number of claims also significantly drop post-AIA. This measure deems a separate study, and the focus of this paper is to estimate the broadest change in quantity and quality of inventions by small entities post-AIA. Therefore, I do not discuss the change in claims further. However, this result stays as an addendum and assists in the mechanism that would be discussed in the next sections.

Now, I discuss the second outcome, the average scaled citations within two years of issue of the patent. From Table 3, we observe results similar to that of the log(patents). Specifically, the estimate of δ reported in columns (3) and (4) of Table 3 indicates a drop in the number of citations within two years of the issue of a patent per patent of about 19.8 percent for small entities post-AIA relative to all entities pre-AIA. This is rather an alarming drop, but this drop is driven by the years which are substantially far away from the AIA's dates, unlike for the results from log(number of patents), which stays robust with a shorter window, different specifications, and additional controls. This compels us to think if this drop in citations is indeed due to the AIA, or due to other changes, unrelated to the legislation. In an event study setup, shown in Table 4, and Figure 5, this becomes clearer, that years 2008 and 2016 in particular are driving this gap. In particular, the yearly coefficients for citations starting from 2009 till 2015 all remain statistically insignificant in Table 4.

While Abrams and Wagner (2013) and Lerner et al. (2015) report a reduction in the number of patents in Canada, they do not find any appreciable difference in the quality, as also measured by citation, for small entities during the Canadian reform. This paper's results show a similar trend in among the U.S. inventors. However, De Rassenfosse (2013)

Table 4: Event study results: log(no. of patents) and avg. scaled citations

	Patents		Citations	
	(1)	(2)	(3)	(4)
SE x 2008	0.0116	-0.0180***	0.1992*	0.3074***
	(0.0072)	(0.0052)	(0.1127)	(0.1115)
SE x 2009	0.0312***	0.0236***	0.1495	0.2285**
	(0.0069)	(0.0050)	(0.1081)	(0.1073)
SE x 2010	0.0214***	0.0302***	-0.0252	0.0466
	(0.0065)	(0.0048)	(0.1045)	(0.1036)
SE x 2011	0.0157***	0.0300***	0.0154	0.0544
	(0.0060)	(0.0045)	(0.1005)	(0.0997)
SE x 2012	0.0035	0.0124***	-0.0195	-0.0012
	(0.0054)	(0.0041)	(0.0913)	(0.0906)
SE x 2014	-0.0323***	-0.0192***	-0.0213	-0.0329
	(0.0051)	(0.0038)	(0.0842)	(0.0836)
SE x 2015	-0.0579***	-0.0353***	-0.1248	-0.1822**
	(0.0055)	(0.0041)	(0.0856)	(0.0851)
SE x 2016	-0.0803***	-0.0546***	-0.3803***	-0.4214***
	(0.0059)	(0.0043)	(0.0868)	(0.0861)
Controls	No	Yes	No	Yes
ID F.E.	Yes	Yes	Yes	Yes
Subcat F.E.	Yes	Yes	Yes	Yes
year	Yes	Yes	Yes	Yes
N	1148250	1146957	1148250	1146957

This table reports the gap between small and large entities for each year before and after the AIA. The number of patents are measured by log(number of patents) and citations are measured by the average scaled 2 citations within 2 year of issue of patent. The estimates are derived from the model $Y_{it} = \beta_0 + \sum_{s\neq 0}(\beta_s \times 1[s = t] \times SE_i) + X_{it}^{'}\beta + \lambda_i + \lambda_t + \varepsilon_{it}$, where i and t denote entity and quarter respectively, X_{it} denotes a range of patent and entity level controls as defined in Table 1, and λ_i, λ_t control for entity and quarter fixed-effects respectively. The entity-quarter level dataset consists of all patents applied between 2008 and 2016. The indicator variable SE takes value 1 if more than 50 percent of the patents for an entity was applied as a small entity, and 0 otherwise. The year 2013 acts as the base year, when the AIA's FITF came into effect.

Columns (1) and (3) report results without and (2) and (4) report with controls. P-values at 1, 5, and 10 percent are denoted by *, **, and ***. Standard errors are clustered at entity-level.

show that trade-off between quantity and quality of patents do exist. This result is not prominent at least when we study patenting activity around the AIA's enactment.

Citations are a tricky and complicated proxy to measure quality of patents. Over time, strategic citations have become prevalent, and overall number of citations have increased per patent (Lerner and Seru, 2022). Because of this, I report results with alternative definitions of citations in Appendix A.6 and A.7. In Appendix A.6, rather than using citations within two years of issue as outcome and then transforming it into its scaled version, I use citations within one year of issue date and show the results in two ways: scaled citations by NBER subcategory and year, and quarterly citations per patent. In Appendix A.7, I separate citations added by the examiner and applicants, and re-calculate the average scaled 2 year citations. All of the exercises show that citations for small entities compared to large entities do not significantly change post-AIA.

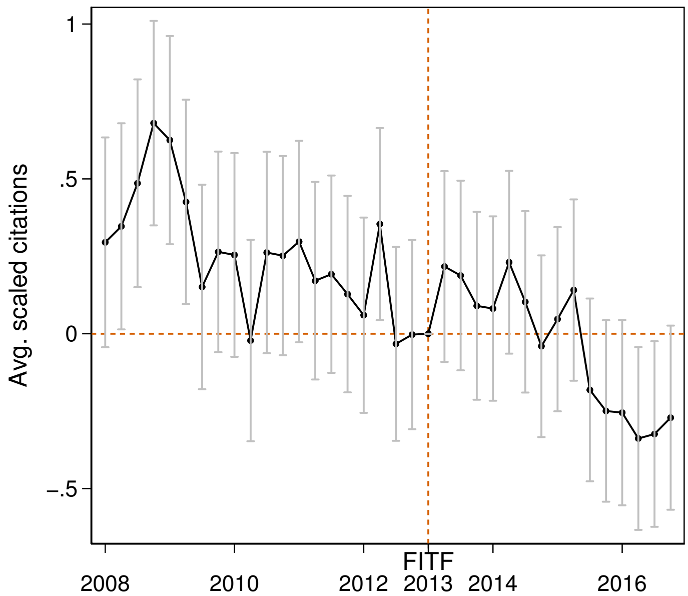


Figure 5: Change in small entities' citations over time
This figure plots the change in 2 yr. scaled citations by small entities for every quarter. Number of citations are transformed into the percentile within NBER subclass and quarter to create scaled citations. The event study specification is reported in Equation 2, and the figure plots the β_s's estimated from $\sum_{s\neq 0} \beta_s \times 1[s = t] \times SE_i$ part of the equation, with the base quarter being the first quarter of 2013, when the AIA's FITF came in force.

The AIA was implemented in stages, as explained in Subsections 2.1 and 2.3. But, these stages are fundamentally different from each other in terms of changes in the benefits and costs of patenting. Therefore, have to be evaluated individually. In the next section — Section 5.2, I use the enactment of AIA as the base quarter and discuss how the results stay the same and do not change qualitatively.

5.2 Narrow window of analyses: 2009-2014

In this section, I compare the results discussed in Section 5.1 for a narrower window of analyses (2009-2014), and with a different comparison quarter — the implementation quarter of the AIA — September 2011, or quarter 3 of 2011.

This section serves two purposes: it illustrates an anticipation for the AIA's first-inventor-to-file rule, and rules out any effect that could be attributed to the recent Supreme Court judgements on *Alice v. CLS Bank* decided on June 2014, and *Bilski v. Kappos* decided on June 2010.[30, 31] This case was on patentable subject matter, or patentability of abstract ideas. Prior to these cases, any "useful results" were patentable, and the Supreme Court did not intervene in the patent litigation process, leaving it to the Federal Circuit. *Alice* held four patents which were challenged by *CLS Bank* on their patentability, and were later invalidated by the Supreme Court. Feng and Williams (2023) and Lemley and Zyontz (2021) study this case in detail. Lemley and Zyontz (2021) study *Alice v. CLS Bank* and the related cases, and their relation to invalidated patents after the case's decision. They find that small and individual inventors were affected adversely due to the case's decision, and the proportion of invalidations increased. Feng and Williams (2023) confirm that *Alice*-like patents were invalidated at a greater rate post-*Alice v. CLS Bank*. This increase in invalidation, as the authors point, were linked to a greater scrutiny by examiners. Activity of Patent Assertion Entities (PAE) and overall litigation fell in *Alice*-like areas which in essence weeded out some of the costs to the patenting system that arises due to excessive litigation. Did this come at a cost of reducing patenting activity of small entities, the central question of this paper? Feng and Williams (2023) find that startups in *Alice*-like industries were not affected due to the Supreme Court decision.

[30] Bilski v. Kappos, 561 U.S. 593 (2010)
[31] Alice Corp. v. CLS Bank Int'l, 573 U.S. 208 (2014)

Table 5 and Figures 6a and 6b report pre and post-AIA averages, and event study results respectively for a shorter window of analyses, 2009-2014, and considers the quarter of the AIA's enactment as the base quarter for comparison — as opposed to the AIA's first-inventor-to-file's quarter. Table 5 serves as a robustness, and shows that mean drop for small entities post-AIA occured over the course of the AIA's implementation, and the coefficients of column 2 are close to that of the main results from column 2 of Table 3. Specifically, we observed a 4.4 percent drop in the longer timeframe, and a 4.2 percent drop for small entities for the shorter timeframe. For citations we observe no statistically significant changes post-AIA. This confirms that the tail years, i.e. 2008, 2015, and 2016 were driving the results.

It is possible that the Great Recession of 2008 induced a different patenting strategy, and thus an increase in citations around 2008-2009, but it is at this point unclear and requires further exploration towards its mechanisms. For the years 2015 and 2016, it is likely that the *Alice* is influencing a drop in citations. Lemley and Zyontz (2021) note that the case on one hand may increase certainty in what can and what cannot be patented, they also have spurred multiple inconsistencies in subsequent patent invalidity judgments, increasing confusion. Because of this, it is possible that we are observing a strategic drop in average citations over all patent categories. Entities are deliberately keeping their inventions secret, and are disclosing only those inventions that are of poor quality, on average.

Figure 6a provides a cleaner view of the widening gap between small and large entities' number of patents post-AIA. While the gap remained insignificant compared to Sept 2011, even after *Bilski v. Kappos* which was decided on June 2010, it starts dropping after the AIA's enactment, showing that at the least, part of the results can be attributed to the AIA rather than only to the Supreme Court decisions. While we do observe a significant change in the number of patents by small entities, the citations do not move away from zero, and the estimates exhibit a greater imprecision, shown by the wide standard errors, as compared to the number of patents' confidence intervals. In the two figures, I mark the two important events of the AIA, its enactment, and the implementation of first-inventor-to-file. The second vertical dotted line compares the results with Figure 4.

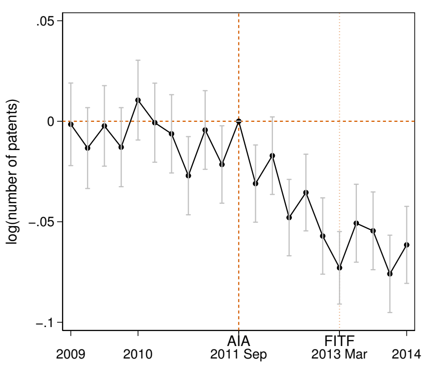
While this paper's objective is to show certain broader results related to small entities' quantity and quality of patents around the AIA, dividing innovating entities into small and large may not be enough. As Lerner and Seru (2022) and Jaffe and De Rassenfosse (2019) note, use of citation to proxy quality of inventions has its own pitfalls, and have been knowingly or unknowingly misused in the literature. Among the distribution of patentees, the entities who file routinely file for a considerably higher number of patents than the average are qualitatively and strategically different from the others. Abrams et al. (2019) show that a group of small entities aggregate, and sell their "weak" patents to non-practicing entities, who in turn use those to invalidate other patents. To disassociate different strategies by patent portfolios of entities, I provide event studies for different subsamples within the distribution of low to heavy patentees in Appendix A.8. I loosely define "low patentees" as the entities who fall between [0-75) percentile in the distribution of number of patents, and "high patentees" as those who fall above the 90^{th} percentile. We do observe interesting behavior among the two types, and in particular, the drop in patenting activity is driven by small entities who are between 90-99 percentile of the distribution of number of patents, as given by Figure A11b. As opposed to this, the patentees having a lower number of patents in their portfolio, on average, spring back after the AIA's first-inventor-to-file's implementation. The top 1 percent are not statistically different from large.

Table 5: Main results with shorter timeframe

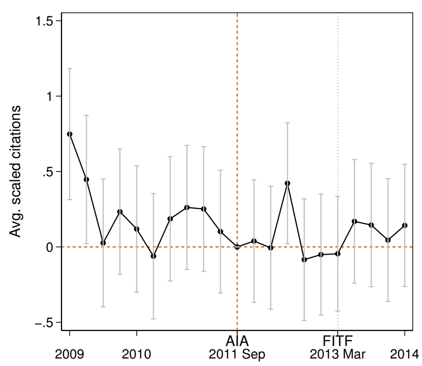
	Patents		Citations	
	(1)	(2)	(3)	(4)
SE x Post	-0.0409***	-0.0380***	-0.0958	-0.0903
	(0.0052)	(0.0040)	(0.0860)	(0.0853)
Controls	No	Yes	No	Yes
Qtr F.E.	Yes	Yes	Yes	Yes
ID F.E.	Yes	Yes	Yes	Yes
Subcat F.E.	Yes	Yes	Yes	Yes
N	457448	456944	457448	456944

This table reports the gap in the patents and citations between small and large entities post-AIA. The number of patents are measured by log(number of patents) and citations are measured by the average scaled citations within 2 year of issue of patent. The estimates are derived from the model $Y_{it} = \delta(Post \times SE_i) + X_{it}^{'}\beta + \lambda_i + \lambda_t + \varepsilon_{it}$, where i and t denote entity and quarter respectively, X_{it} denotes a range of patent and entity level controls as defined in Table 1, and λ_i, λ_t control for entity and quarter fixed-effects respectively. The entity-quarter level dataset consists of all patents applied between 2009 and 2014. The indicator variable SE takes value 1 if more than 50 percent of the patents for an entity was applied as a small entity, and 0 otherwise. $Post$ takes value 1 if the patents were applied on or after the third quarter of 2011 i.e. the AIA's enactment quarter — as opposed to the implementation of first-inventor-to-file rule that was considered in Table 5.

Columns (1) and (3) report results without and (2) and (4) report with controls. P-values at 1, 5, and 10 percent are denoted by *, **, and ***. Standard errors are clustered at entity-level.



(a) Patents



(b) Citation

Figure 6: Narrow window of analyses: 2009-2014

These figures plot changes in log(number of patents) and avg. scaled citations by small entities for a shorter timeline, the first quarter of 2009 to the first quarter of 2014. Also, the base quarter is the AIA's signing quarter (September 2011) rather than the quarter of first-inventor-to-file's enactment (March 2013). The event study specification is reported in Equation 2, and the figure plots the β_s's estimated from $\sum_{s \neq 0} \beta_s \times 1[s = t] \times SE_i$ part of the equation.

5.3 Evidence from the litigation exposure

In this section, I focus on the variation from exposure to litigation, and report results for small entities and all entities separately. In the Appendix A.11, I provide triple difference results, comparing small entities — exposed to litigation — post-AIA.

We observe that in addition to the gap in the number of patents between small and large entities widening post-AIA; within small entities, those operating in patent categories with heavier exposure to litigation do decrease their patenting even more compared to all small entities post-AIA. Specifically, one-standard deviation increase in exposure to litigation results in about 15.7 percent reduction in the number of patents for all entities, and 42 percent among small entities, given by the coefficient of $Exposure \times Post$ from Tables 6 and 7 respectively.

It should also be noted that while on average citations did not significantly change for small entities post-AIA, among those exposed to litigation filed for patents that received lower citations on average, as shown in Figure 7b, and columns 3 and 4 of Table 7. While the number of patents, and citations did drop for all entities, for small entities, we observe an increased effect, and this result is confirmed when I estimate the averages using a triple difference model reported in Appendix A.11.

Post-AIA, we can infer that the entities experiencing heavy exposure to litigation are in a way more "discouraged" to patent in those areas. The mechanism which can explain this result is complex. While the AIA tried to increase certainty, the establishment of Patent Trial and Appeal Board (PTAB) made it easier to oppose patents' claims, and on top of this, the Supreme Court Cases, from *Bilski* to *Alice* exacerbated the "discouraging effect" of filing patents in areas with higher exposure to litigation.

Lemley and Zyontz (2021) report an increase in patent invalidations, and Marco et al. (2017) report a dramatic increase in litigation from 2010. In Figure 7, I separate the pre and post difference in means for each quarter and compare it with the quarter when the AIA's FITF was enacted, only for small entities. We note that the exposed entities were anticipating the change and were lowering their activity till the AIA's passage, and this decline sustains, especially for small entities, post-AIA's first-inventor-to-file. For the number of patents for all entities, we can see a change in the rate of decline, and it nears to zero in the post-AIA period. A drop can also be observed for the citations. When we compare the event studies for only small entities versus all entities, we see that the exposed small entities are the most adversely affected among all. This result is a combination of the Supreme Court cases, and the AIA. At this point, due to the overlapping nature of events, it is difficult to disentangle the effects driven due to the AIA, and due to the cases, and this paper can only inform the average change seen during the period. With a different and nuanced identification strategy, this route can be separately explored in future studies.

Table 6: Exposure to litigation for all entities

	Patents		Citations	
	(1)	(2)	(3)	(4)
Exposure x Post	-0.1165***	-0.1139***	-0.0101***	-0.0093***
	(0.0232)	(0.0226)	(0.0023)	(0.0021)
Exposure	0.2975***	0.2799***	0.0228***	0.0211***
	(0.0557)	(0.0530)	(0.0048)	(0.0045)
Controls	No	Yes	No	Yes
Qtr F.E.	Yes	Yes	Yes	Yes
ID F.E.	Yes	Yes	Yes	Yes
Subcat F.E.	Yes	Yes	Yes	Yes
N	1148250	1146957	1148250	1146957

This table reports difference-in-differences estimates for litigation exposure post-AIA, given by $Exposure \times Post$ for all entities. The number of patents are measured by log(number of patents) and citations are measured by the average scaled 2 citations within 2 year of issue of patent. The estimates are derived from the model $Y_{it} = \delta(Exposure \times Post) + X_{it}^{'}\beta + \lambda_i + \lambda_t + \varepsilon_{it}$, where i and t denote entity and quarter respectively, X_{it} denotes a range of patent and entity level controls as defined in Table 1, and λ_i, λ_t control for entity and quarter fixed-effects respectively. The entity-quarter level dataset consists of all patents applied between 2008 and 2016. The indicator variable SE takes value 1 if more than 50 percent of the patents for an entity was applied as a small entity, and 0 otherwise. $Post$ takes value 1 if the patents were applied on or after the first quarter of 2013 i.e. the implementation of first-inventor-to-file rule in the United States.

Exposure is standardized and is defined in Subsection 3.3.3.

Columns (1) and (3) report results without and (2) and (4) report with controls. P-values at 1, 5, and 10 percent are denoted by *, **, and ***. Standard errors are clustered at entity-level.

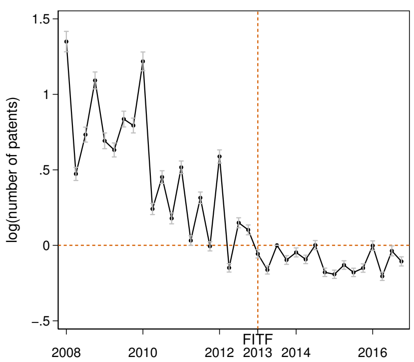
Table 7: Exposure to litigation within small entities

	Patents		Citations	
	(1)	(2)	(3)	(4)
Exposure x Post	-0.3490**	-0.3709**	-0.0514**	-0.0417**
	(0.1674)	(0.1746)	(0.0254)	(0.0207)
Exposure	1.1380***	1.0951***	0.1333***	0.1116***
	(0.2112)	(0.2125)	(0.0287)	(0.0237)
Controls	No	Yes	No	Yes
Qtr F.E.	Yes	Yes	Yes	Yes
ID F.E.	Yes	Yes	Yes	Yes
Subcat F.E.	Yes	Yes	Yes	Yes
N	407900	406918	407900	406918

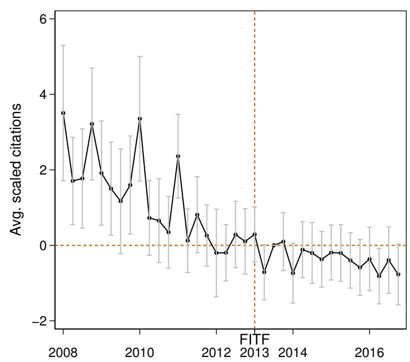
This table reports difference-in-differences estimates for litigation exposure post-AIA, given by $Exposure \times Post$ only for small entities. The number of patents are measured by log(number of patents) and citations are measured by the average scaled 2 citations within 2 year of issue of patent. The estimates are derived from the model $Y_{it} = \delta(Exposure \times Post) + X_{it}^{'}\beta + \lambda_i + \lambda_t + \varepsilon_{it}$, where i and t denote entity and quarter respectively, X_{it} denotes a range of patent and entity level controls as defined in Table 1, and λ_i, λ_t control for entity and quarter fixed-effects respectively. The entity-quarter level dataset consists of all patents applied between 2008 and 2016. The indicator variable SE takes value 1 if more than 50 percent of the patents for an entity was applied as a small entity, and 0 otherwise. $Post$ takes value 1 if the patents were applied on or after the first quarter of 2013 i.e. the implementation of first-inventor-to-file rule in the United States.

Exposure is standardized and is defined in Subsection 3.3.3.

Columns (1) and (3) report results without and (2) and (4) report with controls. P-values at 1, 5, and 10 percent are denoted by *, **, and ***. Standard errors are clustered at entity-level.



(a) Exposure to litigation: Patents



(b) Exposure to litigation: Citations

Figure 7: Small entities' exposure to litigation

These figures plot changes in log(number of patents) and avg. scaled citations for an exposed vs. an under-exposed to litigation only for small entity by quarter. The estimates are derived from the model $Y_{it} = \beta_0 + \sum_{s \neq 0}(\beta_s \times 1[s = t] \times Exposed) + X_{it}^{'}\beta + \lambda_i + \lambda_t + \varepsilon_{it}$, where i and t denote entity and quarter respectively, X_{it} denotes a range of patent and entity level controls as defined in Table 1, and λ_i, λ_t control for entity and quarter fixed-effects respectively. The Appendix Table A5, Figure A14b report these results for all entities.

5.4 Evidence from the Compustat firms

Small entities cover about 30 percent of the patents in the main sample. But, among the publicly traded firms, only about one percent of the patents are from small entities. While this sample enables me to control for a greater number of entities' characteristics, such as assets, R&D expenditure, and the number of employees; I inevitably lose all the individual patentees and a large proportion of unlisted firms. Patents by individuals and small firms are important when studying the AIA because of the results from the Canadian reform, as documented by Lerner et al. (2015), Abrams and Wagner (2013), and Lo and Sutthiphisal (2009). The overarching conclusion all the authors report is that small entities, and particularly individual patentees decrease their patenting activity in Canada, and the authors observe an increased gap between large and small entities after an AIA-like reform in Canada. Publicly traded firms do not adequately represent small entities. Table 8 reports results from two different models given by equations 1 and 3 in rows $SE \times Post$ and $Exposure \times Post$ respectively. In this table, I also add a measure of value of patents derived from the stock market as an alternative measure of patent quality (value) — KPSS value from Kogan et al. (2017).

The results from row $SE \times Post$ show and reiterate that the publicly traded small firms are similar to large firms in terms of their patent strategies, and we do not observe any significant difference between them. They face similar incentives to patent, while the other non-publicly traded small entities, which comprises unlisted small firms and individuals, face a different set of incentives and constraints to patent their invention. From the next row, $Exposure \times Post$, we observe results similar to Table 6 i.e. entities patenting in areas where litigation is prevalent reduce their activity post-AIA. An interesting finding is that the drop in patenting by listed firms is about 3.6 percent, while the same coefficient when is computed for only small entities is about 37 percent. No change post-AIA can be observed for KPSS value.

It should be noted that the publicly listed small entities are still sufficiently large to be listed as compared to the unlisted firms and individuals. They also are significantly less budget constrained than the other small entities. This can be one explanation as to why this sample of small entities do not exhibit a different behavior as compared to larger ones post-AIA.

Table 8: Results for listed firms

	Patents		Citations		KPSS	
	(1)	(2)	(3)	(4)	(5)	(6)
SE x Post	-0.0757	-0.0216	1.0795	1.8093	-0.0118	0.0072
	(0.0579)	(0.0594)	(1.7096)	(1.5000)	(0.0889)	(0.0919)
Exposure x Post	-0.0298***	-0.0356***	-0.0356	-0.0566**	-0.0002	-0.0015
	(0.0099)	(0.0124)	(0.0229)	(0.0276)	(0.0012)	(0.0013)
Controls	No	Yes	No	Yes	No	Yes
Qtr F.E.	Yes	Yes	Yes	Yes	Yes	Yes
ID F.E.	Yes	Yes	Yes	Yes	Yes	Yes
Subcat F.E.	Yes	Yes	Yes	Yes	Yes	Yes
N	18406	17512	18406	17512	18406	17512

This table reports the gap in the patents and citations between *listed* small and large firms, and by exposure to litigation post-AIA. The number of patents are measured by log(number of patents), citations are measured by the average scaled 2 citations within 2 year of issue of patent, and exposure is a standardized weighted average of proportion of litigation by NBER subcategory and quarter where the weights are proportion of patents filed by an entity in a subcategory out of all patents filed by the entity in a quarter. For row *SE x Post* the estimates are derived from the model $Y_{it} = \delta(Post \times SE_i) + X^{'}_{it}\beta + \lambda_i + \lambda_t + \varepsilon_{it}$, and for row *Exposure x Post* the estimates are derived from the model $Y_{it} = \delta(Exposure \times Post) + X^{'}_{it}\beta + \lambda_i + \lambda_t + \varepsilon_{it}$ where i and t denote entity and quarter respectively, X_{it} denotes a range of patent and entity level controls as defined in Table 1, and λ_i, λ_t control for entity and quarter fixed-effects respectively. The entity-quarter level dataset consists of all patents applied between 2008 and 2016. The indicator variable *SE* takes value 1 if more than 50 percent of the patents for an entity was applied as a small entity, and 0 otherwise. *Post* takes value 1 if the patents were applied on or after the first quarter of 2013 i.e. the implementation of first-inventor-to-file rule in the United States. Exposure is standardized and is defined in Subsection 3.3.3.

Columns (1) and (3) report results without and (2) and (4) report with controls. P-values at 1, 5, and 10 percent are denoted by *, **, and ***. Standard errors are clustered at entity-level.

6 Concluding remarks

This paper studies the quantity and quality of inventions by small entities relative to large entities around a recent and substantial change in the patenting rule of the United States, the Leahy-Smith America Invents Act (AIA) of 2011. The AIA brought forth a range of changes in the patenting rules in the United States to put a check on litigation and provide ease in the filing of patents, especially for smaller entities. But, the AIA's Congressional hearings highlighted that small entities' could be disproportionately impacted through the AIA's first-inventor-to-file rule, as they were required to file as quickly as possible upon invention, which adds additional costs to their already constrained resources. Therefore, the AIA's passage required a mandated study to assess the impact of implementing FITF on ssmall entities. Unfortunately, the authors note that the study was premature and warranted further exploration into the questions. Through this paper, I contribute the following: one, I shed light on the innovative activities of a relatively understudied group of entities, small entities before and after the AIA's enactment, acting as an update to Lerner et al. (2015); two, I add clarity to the understanding of the costs and benefits small entities engaged in innovation face and perceive, and its differences with respect to larger entities; and three, I highlight the changes in the trade-offs to file for patents for entities with different sizes as brought-forth by the AIA in the United States.

I study the impact of the AIA in two parts: first, estimate the total change in patenting activity by small entities before and after the AIA, as measured by the number of patents and their citations; and second, estimate the change in patenting activity for entities exposed to litigation. I find that on average, the gap in patents' quantity and quality between small and large entities were widening even before the enactment of AIA. Entities with a greater exposure to litigation were reducing their patenting activity significantly, and among this group, small entities report an even more pronounced gap in patenting. This hints that resource constrained entities are exercising more caution than larger ones. And among them, especially the entities that operate in areas previously plagued by litigation. If certain entities are more cautious to disclose their invention than before, we might expect a drop in follow-on inventions in the years to come. The reasons for a cautious move can be many, and a part of the reasons were discussed by Huang et al. (2020) and Abrams and Wagner (2013). Because an FITF rule recognizes the first filer of an invention as the sole inventor, an entity has to file quickly after invention, but also has to ensure that the patent document is as complete as possible. An incomplete document can cause more harm than good. Second, this might also signal an entity's competitors about the portfolio of inventions that they are developing. Hence, an entity is more likely to ensure greater secrecy as long as they are not fully ready to disclose their invention. This behavior is accentuated by the addition of the post-grant review. A post-grant review can question the validity of any granted patent, and therefore, an entity needs to ensure that questions on validity of their inventions do not arise, or at least are minimized. A drop in follow-on inventions may be deemed as an unintended consequence of the legislation, but this paper's scope may not be adequate to provide evidence on the same. There are multiple aspects of the AIA which remain to be studied, and therefore, determining the AIA to be singularly "good" or "bad" may be premature, and such a conclusion, if drawn, must be thoroughly examined, since a legislation as complex as the AIA is unworthy of a singular label, and requires examinations through multiple lenses.

This paper acts as a first step in analyzing the broadest aspect of the AIA with a focus on small entities. But, a range of questions remain unanswered. We do not know if the AIA indeed resulted in eliminating litigation and "bad patents". While this paper suggests that the additional support small and micro entities received was insufficient to counter the increase in resource requirements the AIA invoked, how was the support used by these entities? Also, if the entities patent at a lower rate than before post-AIA, are they also inventing at a lower rate or are they inventing at the same rate but keeping those secret, and later secretly engaging in licensing deals with large manufacturers? In the years to come, an examination of the follow-on patents are required to assess if the AIA's blanket changes in the patenting rules were too harsh and excessively dampened disclosure of invention or if it hit the sweet spot in dampening the costs to patenting arising from litigation and "bad patents" and simultaneously proliferating invention disclosure through patents.

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A Appendix

A.1 Patent applications without assignee IDs

The Subsection 3.2 describes the sample selection this paper. In the second paragraph, I end up with 5,087,133 patent applications, with or without entity identifiers. Among these, 1,832,053 did not have either assignee ID or inventor ID from the Patentsview database. In this section of the Appendix, I argue that the patent application without the IDs is not a systematic error of the disambiguation algorithm. Rather, most of these patent applications are a derivative of another patent application already considered in the main sample. 87 percent of the patent applications which do not have an ID are either connected to the patent applications in the main sample through a parent or a child application. Among the 87 percent, 99.2 percent are either PCT or provisional applications. These applications are not examined if they are not converted into a non-provisional application within a given time. They either end up being abandoned or are marked as "pending" throughout their life in the USPTO patent database. A summary of the types of patent applications among those who do not have an ID is provided in Table A1. Each row reports an application type, and if they are connected to the main sample through parent or child applications. The main concern here is the utility patents, which amount to 13,203 patent applications. Utility patents may end up being examined, but the other patents will not, and therefore do not pose a threat to the main results of the paper. The proportion of utility patents out of the total missing is minuscule and will not disturb the estimates even if they were in the main sample.

Table A1: Patent applications with missing IDs

Patent application type	Connected	Not-connected	Total
Utility	9,080	4,123	13,203
PCT	901,286	209,622	1,110,908
Provisional	679,304	25,644	704,948
Re-issue	2,146	24	2,170
Re-examination	815	7	822
Missing	1	1	2
Total	1,592,632	239,421	1,832,053

A.2 Results from a balanced panel

In Subsection 3.2, I briefly mention a concern regarding selection that may arise from the use of an unbalanced panel. While the sample is intrinsically not unbalanced, because I do observe all the patents each entity files for each quarter and there are no missing observations for any particular quarter for a given entity; in a definitional sense of an unbalanced panel, the main sample *is* unbalanced.

Different entities may find it favorable to choose between the pre and post-AIA periods to file their patent application which may relate to their objectives and characteristics. If the entities in the groups small and large for the before and after periods are vastly different, estimates showing the change in their patenting activity before and after the AIA may also contain bias. One way to tackle the issue would be to control for enough of the varying entity characteristics which explain their choice between the two periods if any. If we assume that the control variables adequately capture their strategies, the estimates would be consistent. This is one reason why I estimate the model with different samples and variables.

Another way to tackle this is to force the unbalanced panel to be balanced. In the quarters when an entity did not file for patent applications, I put zeroes in the number of patents and citations column. The problem with arranging the data in this way is that it adds zeroes to the time period when an entity was not established. This changes the effect size, and because small entities on average appear at significantly lower rates than larger ones, their representation is negatively skewed.

Table A2 reports results when a balanced panel is forced. We observe that small entities still file for patents at a lower rate than larger ones. Their citations per patent though reported to be significantly higher, is still near zero and not economically significant.

Table A2: Results from a balanced panel

	Patents		Citations	
	(1)	(2)	(3)	(4)
SE x Post	0.4876***	-0.1664***	0.2382**	-0.1958*
	(0.0832)	(0.0395)	(0.0951)	(0.1052)
Individual	-5.1391***		-2.7560***	
	(0.0415)		(0.0475)	
Government	5.3494***		0.0816	
	(0.7196)		(0.8222)	
Constant	6.3705***	2.4955***	3.1808***	1.1168***
	(0.0361)	(0.0101)	(0.0412)	(0.0268)
Qtr F.E.	Yes	Yes	Yes	Yes
ID F.E.	No	Yes	No	Yes
N	1147959	1148250	1147959	1148250

This table reports results from a balanced entity-quarter panel. Entities may or may not file for patent applications in all quarters. I add zeroes to the quarters where the entity did not file for patents, as opposed to the sample in Table 3 which does not have zeroes. The later method is prevalent in the literature. Rest of the descriptions of this table is similar to Table 3.

Columns (1) and (3) report results without and (2) and (4) report with controls. P-values at 1, 5, and 10 percent are denoted by *, **, and ***. Standard errors are clustered at entity-level.

A.3 Results from Poisson regression

Table A3: Results from Poisson model

	Patents		Citations	
	(1)	(2)	(3)	(4)
SE x Post	-0.1117***	-0.0436**	-0.4202***	-0.3433***
	(0.0221)	(0.0195)	(0.1043)	(0.1173)
Controls	No	Yes	No	Yes
Qtr F.E.	Yes	Yes	Yes	Yes
ID F.E.	Yes	Yes	Yes	Yes
Subcat F.E.	Yes	Yes	Yes	Yes
N	1148250	1146957	384502	384147

This table reports results from Poisson regression model for the outcomes number of patents, and number of citations' gaps between small and large entities post-AIA. The estimates are derived from the model $Y_{it} = \delta(Post \times SE_i) + X_{it}^{'}\beta + \lambda_i + \lambda_t + \varepsilon_{it}$, where i and t denote entity and quarter respectively, X_{it} denotes a range of patent and entity level controls as defined in Table 1, and λ_i, λ_t control for entity and quarter fixed-effects respectively. The entity-quarter level dataset consists of all patents applied between 2008 and 2016. The indicator variable SE takes value 1 if more than 50 percent of the patents for an entity was applied as a small entity, and 0 otherwise. $Post$ takes value 1 if the patents were applied on or after the first quarter of 2013 i.e. the implementation of first-inventor-to-file rule in the United States.

Columns (1) and (3) report results without and (2) and (4) report with controls. P-values at 1, 5, and 10 percent are denoted by *, **, and ***. Standard errors are clustered at entity-level.

A.4 Alternate definition of small entities

An entity can file as a small or undiscounted entity. The reason for this choice is unclear. The main set of results assumes an entity to be small if it is ever claimed to be small. But, it is possible that an entity grew over time to be large and be misrepresented as a small entity because of the assumption. Therefore, Table A4 reports results when the variable small entity is not an indicator variable. Rather, it is the proportion of times an entity claimed to be small out of total patents filed for that quarter. This value ranges between 0 and 1 and is a continuous measure of an entity being small and large in each quarter.

I re-estimate Equation 1 considering the proportion of small entity for each quarter in place of the indicator variable small entity. The coefficients of SE $\times$ Post do not qualitatively change compared to Table 3.

Table A4: Results from an alternate definition of small entities

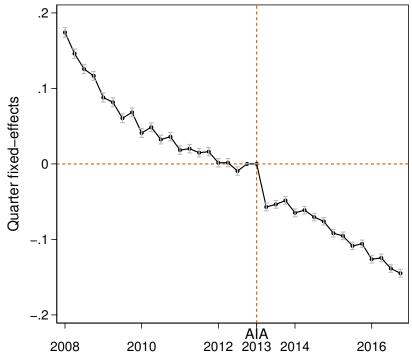
	Patents		Citations	
	(1)	(2)	(3)	(4)
SE x Post	-0.0730***	-0.0416***	-0.1835***	-0.2760***
	(0.0047)	(0.0034)	(0.0640)	(0.0635)
Controls	No	Yes	No	Yes
Qtr F.E.	Yes	Yes	Yes	Yes
ID F.E.	Yes	Yes	Yes	Yes
Subcat F.E.	Yes	Yes	Yes	Yes
N	1148250	1146957	1148250	1146957

This table reports results from the model similar to the main results, as reported in Table 3 but with an alternative definition of small entities. Here *SE* ranges from 0 to 1, and the proportion is calculated as the number of patents filed as a small entity over the total number of patents the entity is involved in inventing. For simplicity, the definition used in the Table 3 *SE* takes value 1 if more than 50 percent of the patents for an entity was applied as a small entity, and 0 otherwise. All other parts of the model is same as in Table 3.

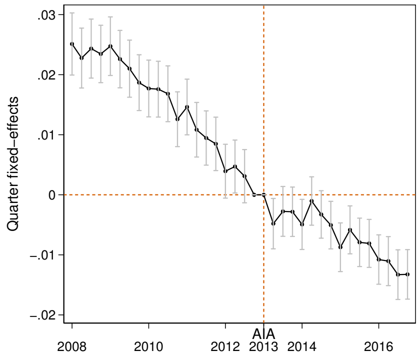
Columns (1) and (3) report results without and (2) and (4) report with controls. P-values at 1, 5, and 10 percent are denoted by *, **, and ***. Standard errors are clustered at entity-level.

A.5 Quarter fixed-effects from the main table

Figures A8a and A8b plot the quarter fixed-effects from Table 3.



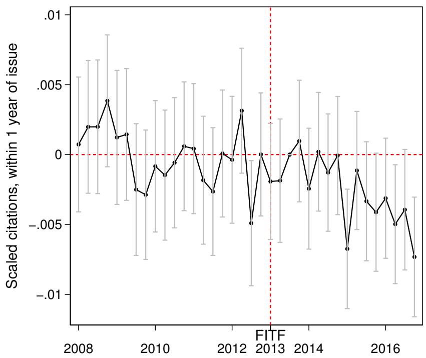
(a) Number of patents' quarter fixed-effects



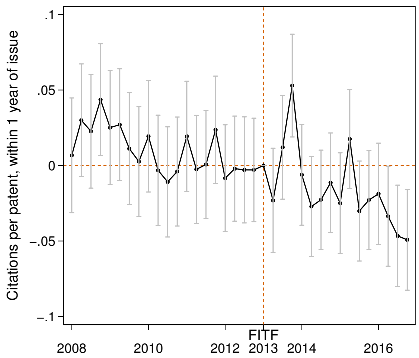
(b) Number of citations' quarter fixed-effects

Figure A8: Quarter fixed effects from Table 3

A.6 Number of citations within one year of issue



(a) Change in small entities' 1 yr. scaled citations over time

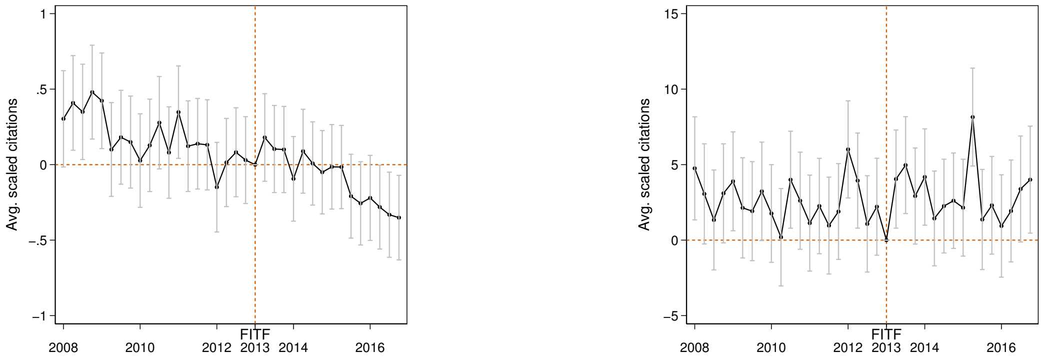


(b) Change in small entities' number of 1 yr. citations per patent over time

Figure A9: Citations within a year of issue, scaled and absolute measure

A.7 Results from applicants' and examiners' citation additions

The Figures A10 report results by separating examiner and applicant added citations. In the main paper, I use total citations and convert it into percentiles for each NBER subclass and year combination. While the examiner and applicant added citations are not different from zero, before and after the AIA, an interesting finding is an increase in applicant added citations, while a drop in examiner added citations, and the estimates are statistically significant towards the later part of the post-AIA period. The effects are prominent after 2014, which may indicate an effect that could be attributed to *Alice*. Examiner added citations are stronger predictor of patent value compared to the applicant added citations (Hegde and Sampat, 2009). Since *Alice* questions the validity of abstract patents, entities are citing granted patents at a greater rate to prove their own patents' validity.

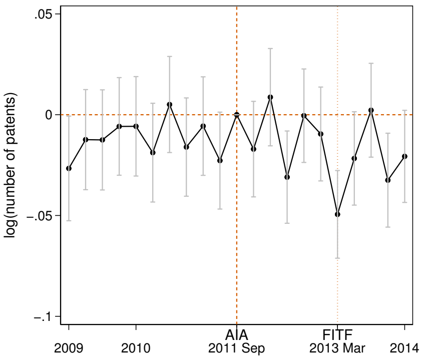


(a) Change in small entities' scaled citations (examiner added)

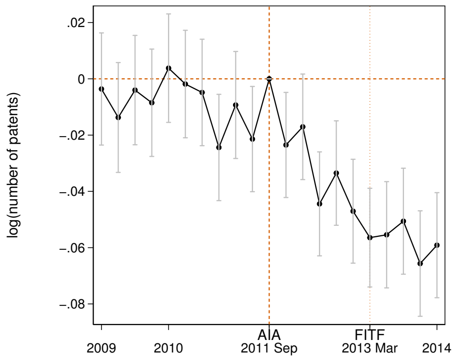
(b) Change in small entities' scaled citations (applicant added)

Figure A10: Citations, examiner and applicant added

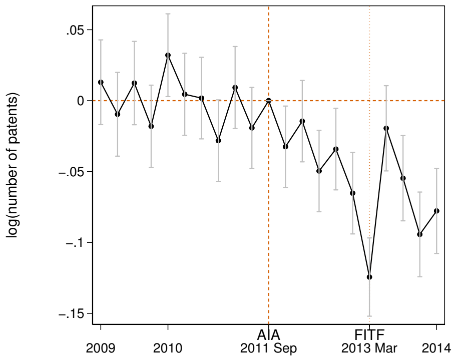
A.8 Patentees' subsample



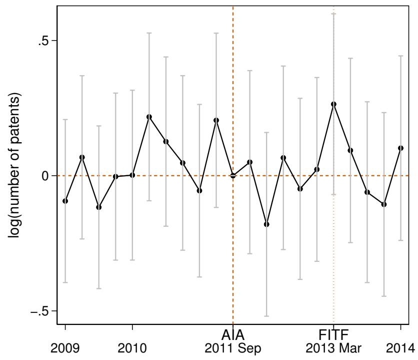
(a) [0-75) percentile



(b) [0-99) percentile



(c) [75-100] percentile

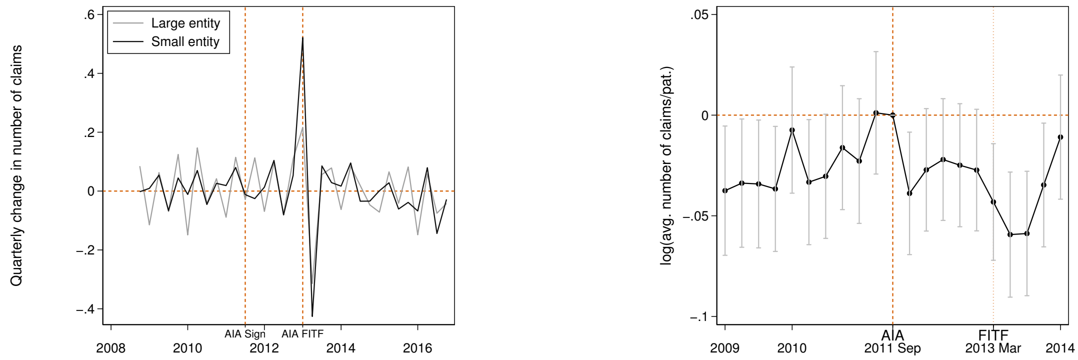


(d) [99-100] percentile

Figure A11: Subsamples by light and heavy patentees

A.9 Results considering from number of claims as outcome

Figure A12 reports changes in log(number of claims per patent) post-AIA for small entities. When we contrast and compare Figure 6a with Figure A12, we observe similar outcomes, which shows that not only did the number of patents drop post-AIA, they also became shorter in length, scope, and its use. This requires a separate study, and therefore is not part of the main paper.



(a) Claims, Quarter-on-quarter

(b) Change in log(average claims per patent)

Figure A12: Change in small entities' claims

Figure (a) reports the change in number of claims quarter-on-quarter for small and large entities. Figure (b) plots the change in log(number of claims) by small entities between 2009 and 2014. The base quarter is the AIA's signing quarter (September 2011). The event study specification is reported in Equation 2, and the figure plots the β_s's estimated from $\sum_{s\neq 0} \beta_s \times 1[s = t] \times SE_i$ part of the equation.

A.10 Cases filed per month

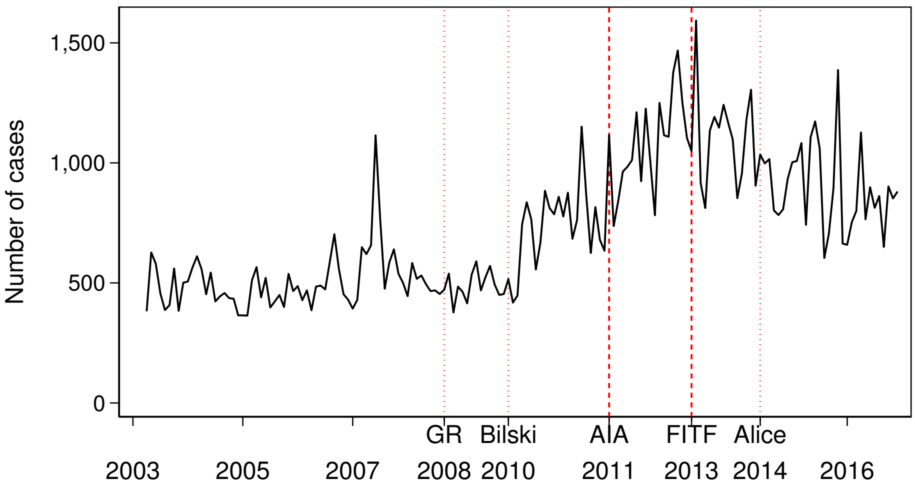
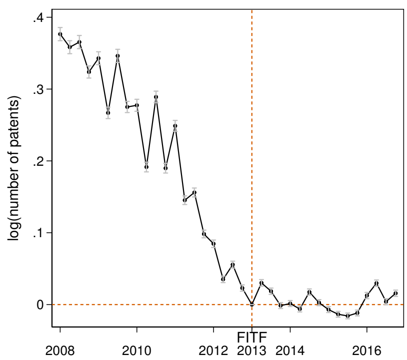


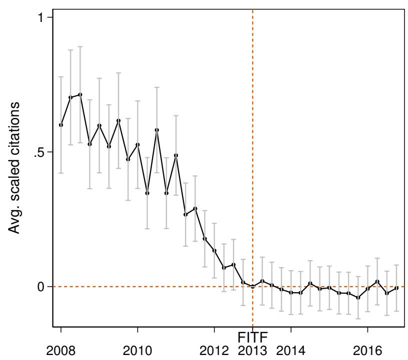
Figure A13: Court cases filed each month

This figure plots total cases filed at the District Courts each month from 2003 through 2017. The important Supreme Court cases are marked: *Bilski v. Kappos* (Nov 2009), and *Alice v. CLS Bank* (June 2014)

A.11 Litigation exposure for all entities



(a) Exposure to litigation: Patents



(b) Exposure to litigation: Citations

Figure A14: Exposure to litigation: all entities

Table A5: DDD estimates for exposure to litigation

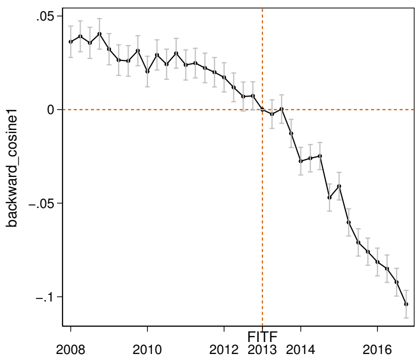
	Patents		Citations	
	(1)	(2)	(3)	(4)
SE x Exposure x Post	-0.2340***	-0.2673***	-0.0417***	-0.0333***
	(0.0068)	(0.0049)	(0.0027)	(0.0027)
SE x Exposure	0.8165***	0.7966***	0.1092***	0.0915***
	(0.0066)	(0.0048)	(0.0027)	(0.0026)
Exposure x Post	-0.0971***	-0.0945***	-0.0074***	-0.0072***
	(0.0011)	(0.0008)	(0.0004)	(0.0004)
SE x Post	-0.1189***	-0.1048***	-0.0179***	-0.0200***
	(0.0030)	(0.0022)	(0.0012)	(0.0012)
Exposure	0.2481***	0.2327***	0.0164***	0.0158***
	(0.0015)	(0.0011)	(0.0006)	(0.0006)
Controls	No	Yes	No	Yes
Qtr F.E.	Yes	Yes	Yes	Yes
ID F.E.	Yes	Yes	Yes	Yes
Subcat F.E.	Yes	Yes	Yes	Yes
N	1148250	1146957	1148250	1146957

This table reports triple difference estimates for small entities — exposed to litigation — post-AIA, given by $SE \times Exposure \times Post$. The number of patents are measured by log(number of patents) and citations are measured by the average scaled citations within 2 year of issue of patent. The estimates are derived from the model $Y_{it} = \delta_1(SE_i \times Exposure \times Post) + \delta_2(SE_i \times Exposure) + \delta_3(Exposure \times Post) + \delta_1(SE_i \times Post) + X_{it}^{'}\beta + \lambda_i + \lambda_t + \varepsilon_{it}$, where i and t denote entity and quarter respectively, X_{it} denotes a range of patent and entity level controls as defined in Table 1, and λ_i, λ_t control for entity and quarter fixed-effects respectively. The entity-quarter level dataset consists of all patents applied between 2008 and 2016. The indicator variable SE takes value 1 if more than 50 percent of the patents for an entity was applied as a small entity, and 0 otherwise. *Post* takes value 1 if the patents were applied on or after the first quarter of 2013 i.e. the implementation of first-inventor-to-file rule in the United States.

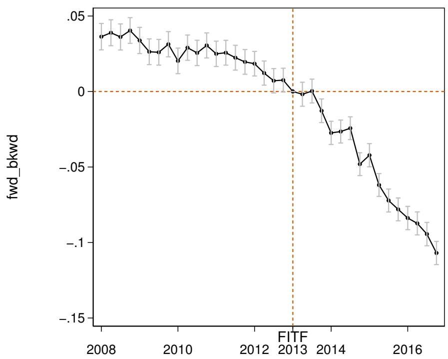
Exposure is standardized and is defined in Subsection 3.3.3.

Columns (1) and (3) report results without and (2) and (4) report with controls. P-values at 1, 5, and 10 percent are denoted by *, **, and ***. Standard errors are clustered at entity-level.

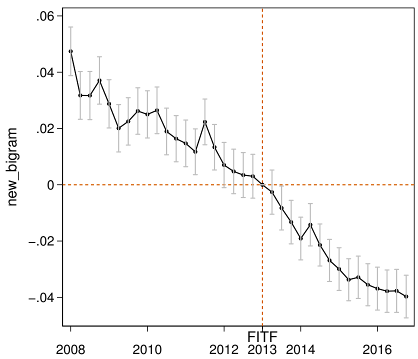
A.12 Measures from Arts et al. (2021)



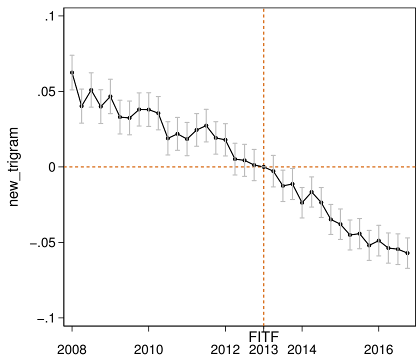
(a) 1-backward cosine



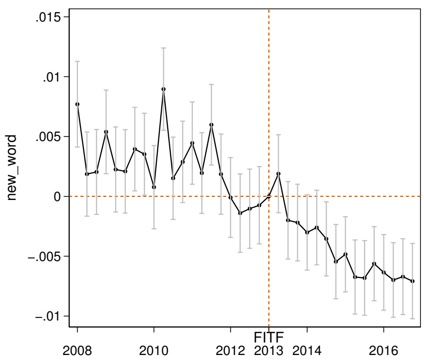
(b) Forward/backward cosine



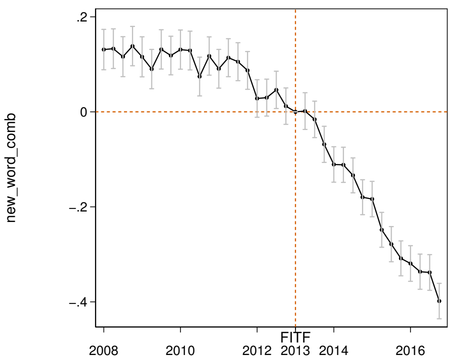
(c) New bigram



(d) New trigram

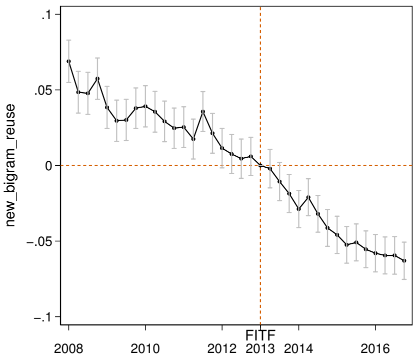


(e) New word

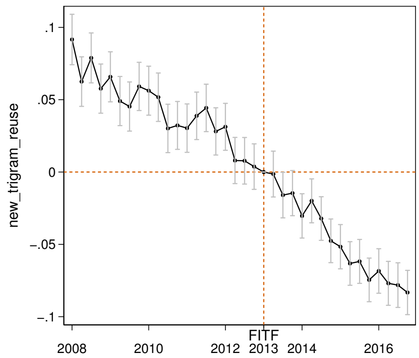


(f) New word comb

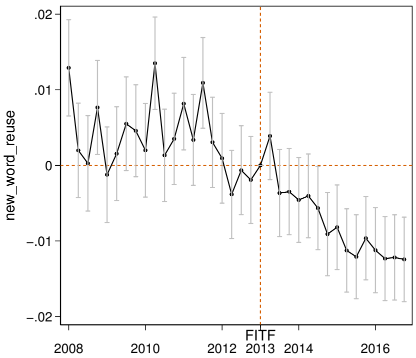
Figure A15: Arts et al. (2021) patent text measures panel (A)



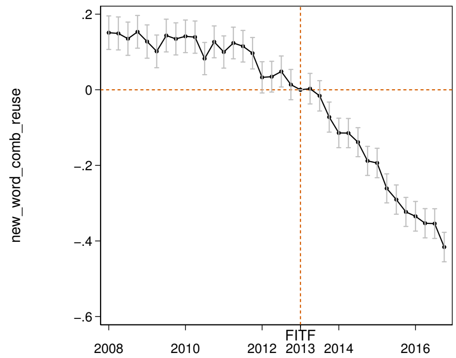
(a) New bigram reuse



(b) New trigram reuse



(c) New word reuse



(d) New word comb reuse

Figure A16: Arts et al. (2021) patent text measures panel (B)