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漫画海賊版が正規版売上に与える影響

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### Keio University



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#### 【要旨】

ネット上の漫画の海賊版が正規版の売り上げに与える影響を推定した。出版社が共同で行った 漫画海賊版の削除プロジェクトを自然実験としたDIDと、海賊版サイトの数を数えた上でのパネ ルデータによる分析の二つの方法を用いた。両者の結果は一致しており、海賊版は連載中の作 品の売上を減らすが、完結した作品の売上を増やす効果があった。完結作品の売上が増えるの はすでに終わった作品を読者に思い出させる効果があるためではないかと考えられる。

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### The Effects of Internet Book Piracy: Case of Comics

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#### Abstract

In this study, the effects of internet book piracy in the case of the Japanese comic book market were examined using direct measurement of product level piracy ratio and a massive deletion project as a natural experiment. Total effect of the piracy is negative to the legitimate sales, but panel regression and difference-in-difference analysis consistently indicated that the effect of piracy is heterogeneous: piracy decreased the legitimate sales of ongoing comics, whereas increased the legitimate sales of completed comics. The latter result is interpreted as follows: piracy reminds consumers of past comics and stimulates sales in that market.

Keywords: copyright; piracy; internet; e-book; comic; Japan JEL: D12,L82,M3,O34

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#### 1. Introduction

Whether piracy reduces legitimate sales has been a hot issue since the internet became the common infrastructure and provided users with new piracy options such as file sharing. Natural intuition tells us that piracy displaces legitimate sales because consumers can enjoy the same product without paying a fee. However, if piracy has some kind of advertisement effect such as sampling, network effect, or marketenlarging through word-of-mouth, and that overrides the displacement effect, legitimate sales may not be decreased (Takeyama, 1994; Peitz & Waelbroeck, 2008). Since copyright theory indicates that both the displacement effect and advertisement effect are possible, the question is left for empirical analysis.

There have been many econometric studies on the impact of piracy in the recent two decades with respect to the music and motion picture industries. Accumulated studies are surveyed in Liebowitz (2006), Oberholzer and Strumpf (2010), and Danaher, Smith and Telang (2013). However, there are very limited econometric studies on piracy in the book industry. One of the reasons that piracy of books has been limited compared with music and movies is because digitization of books lagged behind music and movies. Additionally, e-book readers such as Kindles are specialized devices, not general purpose devices like PCs, making them relatively free from piracy. Piracy can be suppressed on specialized devices more easily due to the limited software working on it. Nevertheless, digitization of books has momentum in this decade to push up the share of e-books to over 40% of publishers' total net revenue in the US and over 10% of sales in Japan in 2014.<sup>1</sup> As people get accustomed to reading e-books on various devices, including general purpose devices such as PCs or smartphones, piracy of books will become a serious concern for publishers in the near future.

The intent of this paper is to examine the displacement and/or advertisement effect of piracy in the case of comics in the Japanese book industry. There are three reasons we chose comics in Japan as a research target for book piracy. Firstly, piracy in Japanese comics is already very common compared with other book categories. One can find pirated illegal copies of almost all popular comics on the internet, and, as soon as a new volume is released from publishers, a pirated copy is uploaded onto piracy sites very quickly. Japanese comic publishers belong to an anti-piracy association named CODA (Content Overseas Distribution Association), which insists that loss caused by overseas piracy is estimated to be double of the overseas legal revenue in 2014.

Secondly, CODA carried out a massive deletion project from July of 2015 till March

 $<sup>^1\,</sup>$  Source: American Association of Publishers for US and Zenkoku Shuppan Kyoukai for Japan

of 2016, which provided us with a natural experiment to test the impact of piracy. Publishers chose some of their comic titles as the target of this project, and CODA sent repeated warnings to piracy sites to delete illegal files from July of 2015. This measure resulted in the decrease of illegal files of target comics to a small but significant extent. If we compare the sales of target comics (treatment group) with non-target comics (control group), we can evaluate the effect of this deletion project. In other words, we can apply difference-in-difference method to the comics market piracy case.

Thirdly, Japanese comics consist of multiple volumes, which reduces the endogeneity problem. Since the early volumes of the series were published several years ago, consumers know the style and quality of the comic very well. Thus, we can assume that the popularity of the comics is established and fixed during the research periods. If this assumption holds, fixed effect term of panel regression removes the effect of the change of comics' popularity, which is the source of the endogeneity problem annoying piracy researchers.

We analyzed 3,360 volumes of 484 comic titles for eight months and applied ordinary panel regression and a difference-in-difference model. Both approaches consistently showed that the effects of piracy on legitimate sales are heterogeneous depending on whether the comics are ongoing or completed. Piracy decreases sales of ongoing comics, but it increases sales of completed comics. To put this another way, displacement effect is dominant for ongoing comics, and advertisement effect is dominant for completed comics. Since completed comics series have already ended, and publishers no longer do any promotion for them, consumers almost forget completed comics. We can interpret that piracy *reminds* consumers of past comics and stimulates sales, just like "discovery" of the music title by streaming service (Aguiar and Martens(2016)) or by new album release (Hendrick and Sorensen(2009))

The remainder of the paper is organized as follows. Section II contains a brief summary of the literature of studies on piracy, and, in section III, the Japanese comic industry is described breifly. In sections IV and V, we explain the data collection procedure and the model specifications, respectively. Section VI contains a report of the results of panel regression using direct measurement of piracy, while section VII features the result of difference-in-difference analysis. Section VIII consists of the summary and discussion.

#### 2. Literature

Early studies on the effects of piracy exploited internet penetration ratio as a proxy of piracy (Liebowitz, 2008; Zentner, 2005). Although this ratio could be measure of piracy at that time, it is not a good measure nowadays because there are many entertainment services on the internet like online music services with fixed fees, YouTube videos, mobile games on smartphone, communication in SNS etc. All these entertainment could crowd out usage of CD and DVD because of consumers' budget and time constraints, thus sales of these old medias may decrease according to the internet penetration even if there is no piracy.

There are two effective approach about the effect of piracy. First approach is to measure each product's piracy ratio directly and look at the correlation between piracy and sales. In this approach, endogeneity is a key issue to be solved because, when the title becomes popular, both sales and piracy simultaneously increase, resulting in a spurious positive correlation between sales and piracy. Another approach is to exploit an exogenous shock as a natural experiment to estimate the causality of piracy. If an exogenous shock, such as legal enforcement, occurs to part of the sample, we can compare the consequences between the sample with the shock (treatment) and the sample without the shock (control).

The results of these two approaches, unfortunately, seem to be mixed so far. Studies using a product level piracy ratio sometimes showed that piracy didn't reduce legitimate sales. Well known paper of Oberholzer-Gee and Strumpf (2007) exploited file sharing data from a P2P server and found that piracy did not reduce music CD sales, using a German school holiday as an instrument variable, although this paper has been criticized with regard to data handling error.<sup>2</sup> Martikainen (2011), Smith and Telang (2009) also measured the piracy ratio directly and reported that piracy did not harm legitimate sales.

Studies using exogenous shock as a natural experiment, on the other hand, tend to indicate that piracy is harmful to legitimate sales. Danaher, Smith, Telang, and Chen (2014) reported that French anti-piracy law, Hadopi (the so-called three strikes law), increased French iTunes music sales by 22-25%. This was shown via difference-indifference method using music sales of other EU countries as the control variables. Adermon and Liang (2014) reported similar results in the case of Sweden's copyright law enforcement in 2009, and Danaher and Smith (2014) found the revenue of major

<sup>&</sup>lt;sup>2</sup> This well-known paper caused long controversy. See Liebowitz(2016) and Oberholzer-Gee and Strumpf (2016)

movie studios increased after the shutdown of famous cyber locker site, MegaUpload. These natural experiments suggest the displacement effect is dominant against piracy on the internet.

In the case of the book industry, to our knowledge, there is no study that either directly measures piracy ratio or exploits a natural experiment. One excellent econometric study is a field experiment done by Hardy, Krawczyk, and Tyrowicz (2014). They chose 94 pairs of similar books, from which one was randomly assigned to be the treatment and the other the control. They intensively deleted unauthorized copies of treatment books during one year, leaving control books to be pirated freely. Comparison of the sales of pairs showed that removal of unauthorized copies had no bearing on legitimate sales. Other literature on the effects of book piracy on the internet is anecdotal. For example, Hilton (2011) reported ten weeks free exposure of eight books online resulted in a 26% increase in legal sales. O'Leary (2009) also claimed there was little connection between file sharing and book sales after putting eight books on free distribution. In spite of this anecdotal evidence, in each case the sample size was too small to evaluate statistically

This study is the first report to examine the effect of internet piracy on the book industry by using a large sample — albeit one that is limited to the specific category of comic books — and applying both a direct measuring approach and a natural experiment approach. We obtained consistent results with these two approaches, which have often shown opposite results in the case of the music and movie industries.

#### 3. Industry Overview

The Japanese comics industry consists of comic books and comic magazines, the total sales of which were 4.046 billion dollars in 2014. Comic magazines consists of short epiosods of 10 to 20 different comic titles and regularly published weekly or monthly, whereas comic book is a collection of approximately a dozen of episodes of a specific title and published quarterly or yearly. Of these two categories, we focus on comic books because their total sales is three times larger than magazines and most pirated files on the internet are copies of comic books, not of magazines.

Comic books' corresponding category in the US is comic and graphic novels. The size of the comic book market of these two countres is shown in Table 1 for comparison. In 2014, the total sale of comic books in Japan was 2.853 billion dollars — almost three times more than US sales of 0.935 billion. When we consider the difference of total book market size between Japan and the US, the difference in comics market size is impressive. If we divide the sales of comics by total book market size, we find that the share of comics in the total book market is 36.1% in Japan, as compared to 2.6% in the US. This means the size of the comics market in Japan is, relatively speaking, over ten times bigger than its counterpart in the US. In Japan, approximately one third of all books are comics. When we divide the sales of comics by total population, we obtain per capita expenditure on comics. This figure amounted to \$2.90/person in the US and \$22.40/person in Japan. The latter number increases to \$31.80/person if we include comic magazines. This implies that the propensity of buying comics is ten times higher in Japan than in the US. Comics (called "manga" in Japan) is a major category in the Japanese book industry.

	comic books	all books	share of comics	expenditure per capita
	(million\$)	(million\$)	(%)	(\$/person)
Japan	2,853	7,898	36.1%	<b>22.4</b> (31.8) <sup>*</sup>
US	935	35,995	2.6%	2.9
	(excluding mag	azines)		

Table 1 Size of comics market in Japan and the US (2014)

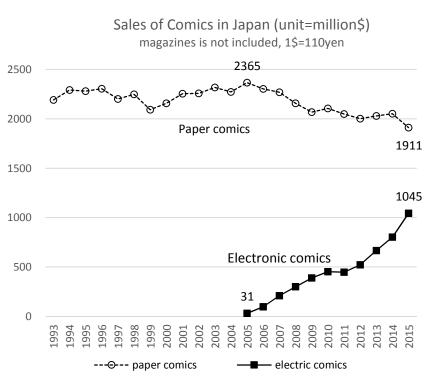
\* when magazine is included

source: ICv2.com

Annual Report and 2014 Global Publishing Statistics, International Publishers Association, monthly report of publications, Zenkoku Shuppan Kyoukai, Japan

The market structure of the comic publishing industry is oligopoly, both in the US and Japan. In the US, two dominant publishers, Marvel Comics and DC Comics, had 100% share in the top 20 ranking and 96% of top 50 ranking in July 2016. There are four major publishers in Japan: Shuei-sha, Kodan-sha, Shougakkan, and KADOKAWA, which held 98% of the top 20 comics ranking in July 2016. (Ranking data is based on ICv2.com for the US and ORICON for Japan).

Figure 1 shows the sale of comics in Japan from 1993 to 2015. Paper comics peaked at 2,365 million dollars in 2005 and have gradually declined to 1,911 million dollars in 2015. Some publishers claimed that this decline was caused at least partly by piracy. However, since sales of electronic comics started to increase from 31 million dollars in 2005 to 1,045 million dollars in 2015, the sales decline of paper comics may be due to just a demand shift from paper to electronics, as is often the case in the digital age. Whether comic piracy harms legal sales or not is still an empirical question.





Unlike the case of the music industry, the main mode of comic piracy is not file sharing, but piracy sites, which consist of cyber locker sites and reach sites. Illegal files are uploaded to the server of cyber locker sites, and reach sites provide users with a search service to find links to the files. Users search for a comic name on the reach site and obtain the link to the file on the cyber locker site. Then, they download the pirated comics from the cyber locker sites. Cyber locker and reach sites are managed by different firms.

The main reason for this division of labor is to avoid legal prosecution. It is illegal to keep pirated files on a server when administrator of the server know that they are pirated files. To put it another way, if administrator who run the cyber locker site are unaware that the file is pirated, they do not face legal punishment. When a copyright holder requests the cyber locker site to delete the file, the site administrator must do so because, at that time, he has learned that it is an illegal file (so-called "notice-and-takedown" policy). Reach sites have no legal obligation because they list only URL links. If the two sites were integrated, it would be clear that the administrator of the integrated site keeps the file on the server with the knowledge that it is illegal, because the site provides a comic search service. Thus, the integrated site would be shut down

Source: monthly report of publications, Zenkoku Shuppan Kyoukai, Japan

by legal authority.

#### 4. Data

In July of 2015, the Content Overseas Distribution Association in Japan (CODA) started the Manga Anime Guardian Project, which requested cyber locker sites to delete the pirated comics files. This was a large project joined by most comic publishers (22 publishers) in Japan. Four major publishers are Shuei-sha, Kodan-sha, Shougakkan, and KADOKAWA, whose total share in the top 30 comic sales in annual ranking during 2010-2014 was almost 95%.

Deleting pirated files is a cat-and-mouse game since deleted files are soon restored. However, this project still had some effect because it covered many publishers and sent requests repeatedly. The actual requesting operation was carried out by a firm hired by CODA, which crawled webs to find pirated files and sent warning messages continuously until the files were deleted. Reportedly, over 90% of piracy files were deleted by these requests, though new files were soon restored with new URLs. In section VII, we show via difference-in-difference method that this project had a small but significant effect on piracy level.

As mentioned in the previous section, there are two types of piracy sites: reach sites and cyber locker sites. Cyber locker sites were the target of this deletion project. For the purpose of our research, however, reach sites were more important than cyber locker sites because, from the view of users, reach sites are the places for they find pirated comics.For example, let us assume that one cyber locker site is shut down, but all reach sites keep active links to pirated files. This assumption is reasonable because reach sites usually list links to multiple locker sites, and/or they can add a new link to another cyber locker site anytime. If the reach sites maintain active links, there is no change to users because they can access pirated files as before. However, if one reach site is shut down, it causes an inconvenience to the users of the site because they have to find another reach site to access pirated comics. Therefore, we focused on the activities of reach sites in our research.

Unfortunately, the number of downloads of each comic title is not open to the public. Thus, we had to use a proxy of the degree of piracy for each comic. For this purpose, we used the number of reach sites that had active links. For example, when there were ten reach sites with active links for comic A and five reach sites for comic B, we assumed that comic A was more pirated than comic B because more users were able to find the link to the pirated files of comic A than comic B. Hardy, Krawczyk, and

Tyrowicz (2014) also used number of available piracy files on the internet as a proxy of the degree of piracy.

There are many reach sites on the internet. Based on hearing survey to publishers and piracy users, we made a list of 25 reach sites. Of these reach sites, we chose 16 major Japanese active reach sites to be researched. "Active" means that the links have been updated recently. "Japanese" means that the pirated comic files have not been translated to non-Japanese language. Since the Manga Anime Guardian Project targeted Japanese sites only, we also limited our research scope to the Japanese site. These 16 Japanese sites are "major" sites in the sense that the pirated files are wellmanaged by the administrators and the sites keep complete volumes of many comics, whereas in other minor sites piracy files are not managed by administrators and often lack complete volumes.<sup>3</sup> The largest reach site of our 16 sites covers 81% of all targeted comics, and the smallest site covers 21%. Since coverage of other minor sites is probably below 20%, we suppose 16 reach sites cover most piracy users.

Almost all comics series consist of multiple volumes. Of these volumes, we chose the first, middle, and most recent two volumes as samples. If the comic consists of ten volumes, for example, we count pirate links of volume 1, 5, 9, and 10, respectively. The reason for this choice is that readers of these volumes have different characteristics. The readers who buy recent volumes are more likely fans of the comic and have purchased previous volumes already, whereas the ones who buy the first or middle volume have noticed the comic recently and started reading. This heterogeneity may influence the effects of piracy. In order to take this heterogeneity into consideration, we count the number of reach sites where the first, middle, and most recent two volumes were linked.

Both sales and number of reach sites are monthly data. Sales data starts in March of 2015 and ends in February of 2016. Counting the number of reach sites began in June of 2015, a month before the massive deletion project started, and ended in January of 2016. We checked the reach site basically the last week of the month. Since there isn't systematic regularity about the timing when the new volumes come out at the reach site, checking the reach sites at the last week of the month will approximately measure the change of links on the reach site.

To make the data structure clear, we show the hypothetical case of a comic in Table 2. This comic is assumed to consist of ten volumes, volumes 9 and 10 of which were

<sup>&</sup>lt;sup>3</sup> Actually 4 of remaning 9 (=25-16) reach sites were almost empy and 3 of them had not updated at least for a year. Last two sites consisted of links uploaded occasionally by users not regularly by admistrators, thus there were many lacking volumes for almost all titles.

published in August and December of 2015, respectively. Volumes 1 and 5 were published several years ago. Distribution of sales of newly published volumes 9 and 10 was skewed because sales were highly concentrated to the initial few months as is often the case of entertainmet goods such as music CD and movie DVD. The number of reach sites that had active links to volume 1 and 5 slightly decreased from six sites in June to approximately five sites after July, thanks to the massive deletion project. The decrease in the number of reach sites was small because of the cat-and-mouse game structure of deletion. Regarding volume 9, links to pirated files do not appear on the reach site in the published month of August, but they appear on five sites a month later. As for volume 10, three reach sites began listing links in the publishing month. This paper's research question is whether the number of reach sites reduced comic sales and whether the massive deletion project increased sales.

Table 2	Data	Structure	(example)
---------	------	-----------	-----------

	year	r 2015						20	16				
	month	3	4	5	6	7	8	9	10	11	12	1	2
	vol1	100	100	100	100	100	400	200	100	100	500	200	100
ممام	vol5	100	100	100	100	100	300	200	100	100	300	100	100
sale	vol9						4000	1000	800	500	400	400	400
	vol10										5000	1500	1000
number	vol1				6	5	5	4	5	5	6	5	
of	vol5				6	5	5	4	5	6	6	5	
reach	vol9						0	5	5	6	6	5	
sites	vol10										3	5	
	↑ deletion start in July												

<sup>↑</sup> counting start in June

Covered in this research are comics published by four major publishers (Shuei-sha, Kodan-sha, Shogakkan, and KADOKAWA). These four publishers listed 322 comic titles as the targets of the massive deletion project. As control data for difference-indifference regression, we added another 162 comic titles published by these four publishers that were not the target of the massive deletion project. Therefore, the total number of comics we counted on the reach sites was 484 (=322+162). Almost 70% of comics are digitized; therefore, sales data of the electronic versions was also available. We treated sales of the paper version and electronic version separately because electronic books and paper books are not interchangeable (Hu & Smith, 2013). Adding the paper versions and electronic versions together, the total sales data to be analyzed increased to 840, 551 of which belonged to the treatment group and 289 of which belonged to the control group. Since we chose four volumes from each comic, the total number of volumes was 840\*4=3360, which is the maximum number of groups of the panel regression. With approximately ten months of data for each volume, as shown in Table 1, the total data size was over 30,000.

Of our 840 comics, 601 were ongoing comics in the research periods, meaning that they ended in 2015 or continued into 2016. The remaining 239 were completed comics that ended until 2014. These categories, ongoing and completed, are important in this paper.

Counting the number of reach sites was done manually rather than via web scraping software because URL expression of links on reach sites changed over time and sometimes became inactive. As a result of the massive deletion project, piracy files on the cyber locker sites were frequently deleted and soon restored. Therefore, the URL links on the reach sites often became inactive or expressed differently. To obtain accurate information, we visited every reach site monthly and checked to see if the links were active or not by clicking the URLs and exiting just before the download process began. In other words, we checked manually 16 reach sites every month with regards to 3,360 volumes.

Book sales data is not shipping-based but actual sales base which was offered by the four publishers for the purpose of academic analysis.

#### 5. Model Specifications

We used two models to examine the effect of piracy on legal sales. The first was ordinary panel regression, which correlated comic sales to the number of reach sites controlling the comics' popularity as fixed effect.

$$\log(Y_{it}) = a_i + b*\log(Site_{it}+1) + c*Z_{it} + e_{it}$$
(1)

Y<sub>it</sub> represents the sales of comics i in month t, whose unit is number of books. Site<sub>it</sub> is the number of reach sites that have active links to comic i in month t. Log transformation was applied assuming nonlinear effect of reach site on the sales. Z<sub>it</sub> is a vector of other exogenous variables that affect the sales of comics, such as media mix, TV promotion, animation etc. The error term e<sub>it</sub> is an idiosyncratic random variable. The coefficient b represents the effect of piracy — either displacement effect (b<0) or advertisement effect (b>=0).

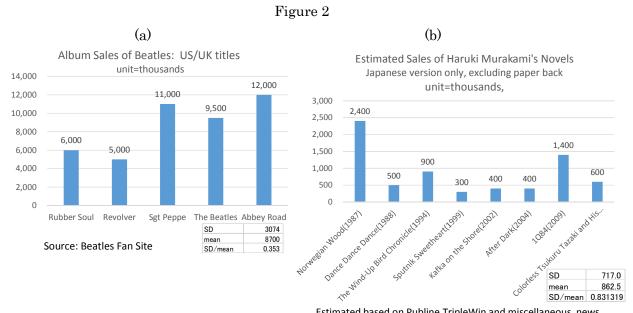
The popularity of individual comics, which is a source of endogeneity, was assumed to be controlled by fixed effect ai. This assumption is valid if popularity was constant within the research periods. The assumption that popularity did not change during the research periods is surely unrealistic and misleading in case of the music industry studies. However, unlike the music industry, there were two reasons to adopt this assumption in our research. First, in the case of comics, the quality of product was already known to consumers because all comics consisted of multiple volumes, and the early volumes were published several years ago.

In the case of the music industry, the quality of CD albums is not known to consumers at the time when the albums are just released. Since the quality of CD albums changes from album to album even if the artist is the same, consumers gather information about the quality of the album through reviews in music magazines, comments on music web sites, word-of-mouth from friends, trial listening etc. If consumers find the quality of the album to be excellent, they begin to purchase CDs or download its pirated files. Thus, in the music industry studies, CD sales and downloading piracy naturally increased simultaneously in the research periods, which inevitably caused an endogeneity problem. That is the reason why most studies on music CD piracy have exploited instrumental variables.

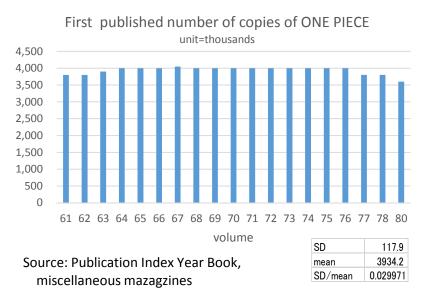
However, in the comic case, early volumes were published several years ago. The median starting year in our data is 2009. Thus, half of the comics in our study started over six years ago. Seventy five percent of comics started before 2012, and 95% started at least one year prior to the research period. What this means is that information about the comics such as characters, storyline, and reviews are already open to public. Consumers can make their subjective evaluations of the comic before publication of new volumes. In other words, consumers know the quality of the comic already because each volume is a part of the same content that started years before. Hence, it is a reasonable assumption that the popularity of the comics is established and constant during the research periods. The main factors varying the sales with time in the research periods are time trends and promotions like TV animation or electronic promotion, which are represented by controlled variable Z.

As examples of this claim I will show three cases of well-known musician, novel writer, and popular comic title. Figure 2(a) is sales of album records of Beatles, one of the most famous music group in the postwar era, and Figure 2(b) is sales of books of novel writer Haruki Murakami, candidate of Novel prize, showing clearly the sales change from album to album and book to book. Figure 2(c) is the sales of comic title

ONE PIECE from volume number 61 to 80, indicating that the sales is very stable. The coefficient of variation (=standard deviance/mean) is 0.353 and 0.831 for records and novels, whereas it is 0.03 for the comic. Theses graphs indicate that it is plausible assumption that the popularity of comic title is established during research periods and can be represented by fixed effect of panel regression.



Estimated based on Publine, TripleWin and miscellaneous news



(c)

The second reason that we assumed fixed effect could control endogeneity is that

the explanatory variable, the number of reach sites, is determined by site administrators and not by consumers. In most research of music piracy, explanatory varible is number of downloads, and both the number of downloads and CD sales are determined by the same consumer; thus, correlation between them is almost certain and exactly simultaneous. In our research, the explanatory variable was not the number of downloads but the number of sites with active links to the comic, and whether reach sites list links to the comic or not is determined by site administrators who do not have information about the changing popularity of the comic among consumers.

How can administrators simultaneously learn about a change in the popularity of comics among readers? Although product price is a signal of demand change in the ordinary market transaction, there is no price of piracy. The price of paper books cannot be a signal because the retail book price is fixed by publishers in Japan, owing to a resale price maintenance regulation. Comic ranking in ORICON is useless because it shows only the top 50 sales weekly; hence, sales of the remaining titles, which make up over 90% of the total, are unknown. If the reach site already had the link to the comic, the administrator could learn the change in popularity by looking at the consumer click count of the link. However, since we are discussing whether the administrator lists the link on the site or not, the site does not have the link yet. There is no systematic way for administrators to simultaneously know the change in popularity of comics they have not listed yet. Therefore, there will be considerable lag for administrators to learn the change in popularity and add new links to their reach sites, which make simultaneous correlations between changes in popularity and the number of reach sites very low or almost zero. If popularity change does not correlate with explanatory variables, we can include it in the error term eit without endogeneity bias.

Considering these two reasons, it is not an unrealistic assumption that fixed effect a<sub>i</sub> can control the difference of the popularity of comics. Therefore, as the first approach, we adopted the ordinary panel regression, assuming that fixed effect a<sub>i</sub> controlled the popularity of the comics.

Although the assumption of constant popularity might be probable, it is not certain. Changes in ongoing comics such as a shift in storyline or the death of a popular character can affect sales even within a research period. Also, exogenous shocks other than Z (tv animation or promotion) may impact popularity. For example, if a big sports event like the summer Olympics is held in the research periods, sports comics may be sold more during the event periods, and administrators may notice this change and quickly added new related links to their sites. Thus, popularity can change during a research period to some extent, which causes an endogeneity problem.

To cope with this problem, as a second approach, we applied standard differencein-difference regression using the massive deletion project that started in July 2015 as a natural experiment.

$$Log(Y_{it}) = a_i + b_0 * DafterJuly_t + b* Deletion_i * DafterJuly_t + c* Z_{it} + e_{it}$$
(2)

DafterJuly<sub>t</sub> is a dummy variable equal to one if t is after July 2015, and Deletion<sub>i</sub> is a dummy variable equal to one if comic i is a target of the massive deletion project. If coefficient b is positive, we can say that the massive deletion project increased the sales of comics because targeted comics were sold more than non-targeted comics after the deletion project, which suggests that piracy decreased legal sales. This regression can control the endogeneity problem because it does not use number of reach sites as an explanatory variable.

To apply this approach, two key variables, DafterJulyt and Deletioni, had to be randomly chosen. We can suppose that DafterJuly was randomly chosen because July was selected based on technical reason of this project, and most consumers and site administrators were unfamiliar with this project before it started. However, Deletioni was not randomly chosen because the target comics of this project was chosen by publishers. It is likely that publishers chose popular comics or comics that they wanted to promote for some reasons. In order to address this bias, we had to choose control comics with similar characteristics to the target comics. We examine this similarity later.

As other exogenous variables  $Z_{it}$ , we adopted the following six variables and monthly dummies.

- *time*: Since all entertainment goods gradually loses popularity after the initial boost, sales of comics also tend to decrease slowly with time. The variable *time*, therefore, is a linear trend variable starting from one to twelve during the research periods in order to remove this general tendency. This is a common linear trend applied to all comics including old volumes which were published years ago.
- *log(trend*<sub>it</sub>): Sales distribution of newly published volumes is quite different from old volumes because sales in the first month are typically extremely high and then decrease very rapidly which is often the case with entertainment goods like CD, DVD movies, and video games. To represent this skewed distribution, we adopted

the variable *log(trend*<sub>it</sub>), trend, which starts from one at the month when the volume was published in the research periods. Log transformation was applied to present the skewed pattern of declining sale. This variable was applied only to volumes published in the research periods — that is, the recent two volumes of ongoing comics.

- *newpub1*<sub>*it*</sub>, *newpub2*<sub>*it*</sub>: When a new volume is published, sales of the first and middle volumes also increase because of the advertisement effect of the new volume. These two variables are dummy variables for this advertisement effect, equal to one after the recent volumes of the comic are published. Newpub1 is equal to one after the second recent volume, and newpub2 is one after the most recent volume.<sup>4</sup> Since average interval of publilication of new volumes is approximately 6 month, some of comics have only newpub2 dummy. These variables were applied to ongoing comics only because completed comics do not have newly published volumes.
- *animeit*: When the comic becomes an animated TV series, the sale of the comic book gains momentum. Variable *animeit* is equal to one if the TV animation series of the comic i was on air at the month t.
- e\_promotion<sub>it</sub>: As for the electronic version of the comic, sales heavily depend on promotion by the electronic shop sites. If the electronic shops put the comic's advertisement on the banner of the top page, the comic sells well. Also, electronic shops sometimes set several comics' prices down by 50% or more as promotion since the marginal cost of supplying the electronic files is almost zero, resulting in a boost in sales. E\_promotion<sub>it</sub> is a dummy variable equal to one if the comic was advertised or discounted on the electronic shopping sites.
- *Monthly dummies*<sub>t</sub>: Monthly dummies were included to remove seasonal variation and temporary monthly common shocks.

Price was not included as a control variable because book price is fixed by publishers in Japan. Retail book stores in Japan are not allowed to change book retail prices bacasuse of resale price maintenance regulation<sup>5</sup>; thus, book prices did not change in the research period, with the exception of promotion in electronic shops. The expected sign was negative for "time" and "log(trend)", and positive for "anime", "newpub1", "newpub2", and "e\_promotion".

<sup>&</sup>lt;sup>4</sup> In case of Table2, newpub1 is equal to one after August 2018 and newpub2 is equal to one after December 2018.

<sup>&</sup>lt;sup>5</sup> Resale price maintenance regulation in Japan obliges resale book store to sell books at the list price set by publishers, and publishers never change the price of the books with a few exceptional promotion of electronic comics which is controlled in our regression by e\_promotion variable.

Descriptive statistics are shown in Table 3. Mean sales were 1,165 books/month, which was far different from their median 70 books/month, indicating that distribution of sales was skewed (skewness = 41.75). Log-transformation of sales made the mean and median almost the same number, 4.2, and the distribution became not skewed (skewness = 0.106). The mean number of reach sites was 7.163, with a distribution similar to normal distribution except for many zero points. One fourth of the number of sites were zero because it takes some time for less popular comics to appear on the reach sites. In that case, the number of sites continues to be zero until pirated files are firstly uploaded on the Internet.

Variable	unit	Obs	Mean	Median	Std Dev	Min	Max
sale	#	31167	1165	70	11353	0	999826
log(sale)		31167	4.202	4.263	2.321	0	13.815
site	#	26880	7.163	8	5.083	0	16
time		40320	6.5	6.5	3.452	1	12
ltrend		14021	0.977	1.099	0.920	0	2.565
newpub1	dummy	20160	0.500	1	0.500	0	1
newpub2	dummy	20160	0.322	0	0.467	0	1
anime	dummy	31167	0.041	0	0.198	0	1
e_promotion	dummy	31167	0.036	0	0.186	0	1

Table 3 Descriptive Statistics

#### 6. Result of Panel Regression

Table 4 shows the result of the panel regression of model (1). Column (1) in Table 2 uses all comics. Before examining the effect of reach sites, let us see other control variables to check the validity of the estimation. Coefficients of time and log(trend) were negative, as expected, which indicates that initial sales rapidly decreased at a rate of power t<sup>-1.242</sup> for newly published volumes after the initial boosts and gradually declined at 3.32% per month as a general tendency. Other exogenous variables, newpub1, newpub2, anime, and e\_promotion were positive, as expected. Most recently published volumes of ongoing comics increased sales of former volumes by 29.1%. TV animation series boosted comic sales by 87.8%, and electronic promotion increased sales by 51.7%. These estimated figures are plausible from the view of Japanese comic business experience according to my interview survey to the four publishers, which supports the validity of this estimation.

The coefficient of reach sites is negative in column (1), indicating that piracy reduced legal sales in the research period. However, when we divide the comics into ongoing comics (column (2)) and completed comics (column (3)), the coefficients show opposite signs, though one of them is insignificant. As for ongoing comics, a 1% increase of reach sites *reduced* sales by 0.102%, whereas the effect on completed comics was not significant and slightly positive. In other words, the displacement effect was dominant in the case of ongoing comics, though some kind of advertisement effect worked in the case of completed comics and cancelled the displacement effect.

	(1)	(2)	(3)
	All comics	Ongoing	Completed
		comics	comics
VARIABLES	Log(sale)	Log(sale)	Log(sale)
log(site+1)	-0.0891***	-0.102***	0.132
	(0.0287)	(0.0305)	(0.104)
time	-0.0332***	-0.0178***	-0.0453***
	(0.00314)	(0.00566)	(0.00383)
log(trend)	-1.242***	-1.294***	
	(0.0637)	(0.0624)	
newpub1	0.0286	-0.0185	
	(0.0339)	(0.0369)	
newpub2	0.291***	0.255***	
-	(0.0263)	(0.0307)	
anime	0.878***	0.879***	1.059***
	(0.0673)	(0.0705)	(0.224)
e_promotion	0.517***	0.483***	0.638***
	(0.0323)	(0.0364)	(0.0753)
Constant	4.976***	5.856***	3.152***
	(0.0625)	(0.0729)	(0.211)
Monthly dummies	included	included	included
Observations	24,562	15,754	7,356
R-squared	0.205	0.232	0.115
Number of group	3,372	2,256	932

Table 4 Basic case: ongoing comics vs completed comics

Robust standard errors in parentheses \*\*\* p<0.01, \*\* p<0.05, \* p<0.1

The reason that advertisement effect appears for completed comics is supposed to be that readers do not remember the completed comics because they ended several years ago, and publishers do not advertise them. Piracy "reminds" consumers of the existence of the completed comics of years ago, and handfuls of downloaders of pirated files find them interesting. A small percentage of them, or those informed by word-ofmouth, purchase the comic. In contrast, ongoing comics are well-known because of new episodes in weekly comic magazines and publishers' advertisements in the book store. Therefore, the "remind effect" of piracy has relatively little effect on ongoing comics. Remind effect is a kind of "backward spillover" (Hendrick and Sorencen(2009)). Hendrick and Sorensen, in the music industry research, found that releasing new album CD caused substantial and permanent increase in sales of the artist's old albums and they interpted this phnenomea as consumers "discovered" the artists. Similarly in the comic industry, consumers are supposed to be reminded of past comics through the pirated free comics

To test the validity of this "remind effect" hypothesis, we divided the data into the first & middle volumes and recent two volumes. As for ongoing comics, publishers and book stores advertise recent volumes in-store more than first and middle volumes. Thus, the benefit of remind effect is minimal in the case of recent volumes of ongoing comics since they are already advertised very well. Relatively speaking, then, the recent volumes suffer more from piracy than the first and middle volumes because recent volumes obtain less benefit from the remind effect.

Regarding completed comics, people who notice the comic by remind effect and have interest in it will start to buy the comics from the first volume, not from the most recent volume, and some readers may stop reading before reaching the final volume.<sup>6</sup> Therefore, the remind effect works more for the first and middle volumes than recent volumes in the case of completed comics. In summary, if the remind effect is working, recent volumes suffer more from piracy for ongoing comics, and first and middle volumes benefit more from piracy when it comes to completed comics.

To investigate this, we ran regressions of columns (2) and (3) of Table 4 dividing the data into recent two volumes and first and middle volumes. Table 5 shows the results. In the case of ongoing comics, the elasticity of the recent two volumes (-0.0967) was two times higher than the first and middle volumes (-0.0397), showing that recent volumes suffer more from piracy than the first and middle volumes. Regarding the completed comics, the coefficient of the first and middle volumes was significant and very high (0.522). These results are consistent with the expected pattern discussed in the previous paragraph, which supports the remind effect hypothesis.

<sup>&</sup>lt;sup>6</sup> It is common fact in the book market that when the book consists of two volumes volume 1 sells more than volume 2. The reason is that there are some readers who buy volume 1 and quit buying volume 2, whereas there is no reader who would buy volume 2 only.

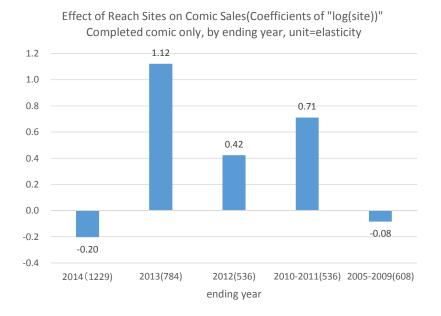
	(1)	(2)	(3)	(4)
	Ongoing	Ongoing	Completed	Completed
	Recent two	First & middle	Recent two	First & middle
	volumes	volumes	volumes	volumes
VARIABLES	Log(sale)	Log(sale)	Log(sale)	Log(sale)
log(site+1)	-0.0967**	-0.0397	-0.0401	0.522***
	(0.0375)	(0.0246)	(0.110)	(0.115)
time		-0.0684***	-0.0529***	-0.0377***
		(0.00466)	(0.00592)	(0.00481)
log(trend)	-1.578***			
	(0.0718)			
newpub1		0.168***		
I		(0.0360)		
newpub2	0.357***	0.386***		
I	(0.0630)	(0.0274)		
anime	0.643***	0.981***	1.167***	0.953***
	(0.155)	(0.0655)	(0.317)	(0.314)
e promotion	0.354***	0.531***	0.609***	0.664***
_1	(0.0660)	(0.0428)	(0.107)	(0.106)
Constant	7.836***	4.670***	3.678***	2.179***
	(0.119)	(0.0589)	(0.221)	(0.247)
Monthly dummies	included	included	× ,	~ /
Observations	6,478	9,276	3,663	3,693
R-squared	0.269	0.214	0.123	0.122
Number of group	1,088	1,168	467	465

Table 5 Comparison of recent two volumes and first & middle volumes

Robust standard errors in parentheses \*\*\* p<0.01, \*\* p<0.05, \* p<0.1

How far into the past does remind effect extend? To answer to this question, we divided the completed comics by ending year and applied the same panel regression. Figure 3 shows the coefficients of log(site+1) depending on the ending year of comics, which is shown along the horizontal line. Regarding periods before 2010, several years were grouped into one category in order to keep the sample size at a similar level (sample size is in parentheses). Figure 3 shows that the elasticities of reach sites were positive for comics that ended in 2010-2013, thus suggesting that the remind effect was effective for this period.

Interestingly, elasticities were almost zero in the near past (in 2014) and long ago (during 2005-2009). The fact that remind effect was ineffective for 2014 can be explained naturally: consumers remembered the comic well because it ended only a year ago. Note that this paper's research period is 2015-2016. If consumers remember the comic well, there is no need for pirated files to stimulate consumers' memories to encourage them to purchase the comic. The reason that remind effect did not work before 2009 is uncertain. One of the possible explanations is that consumers lost interest in the comics of that time period because the style or taste of comics become out of date.



#### Figure 3

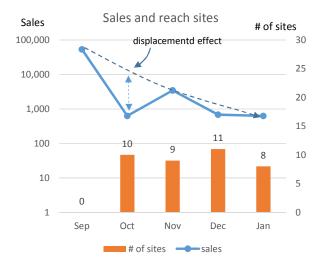
Since studies on the effects of piracy in the book industry are seldom seen, we will show a typical example for illustrative purposes. Figure 4 shows five cases. The horizontal axis is month, and vertical lines are sales for the left axis and number of reach sites for the right axis. Figure 4(a) shows the case of newly published volumes when there was no reach site with links to this comic. Without reach sites, sales decreased continuously after publication. A slight increase in December was caused by the publication of new volume of this comic.

#### Figure 4

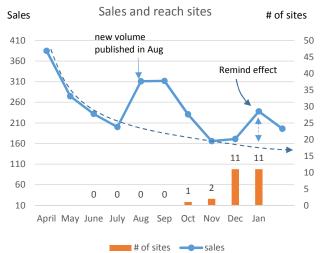
(a) no piracy case: a comic for girls, second recent volume



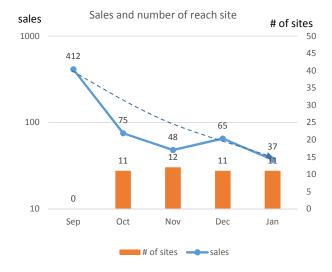
#### (b) Displacement effect case1 a comic for girls, most recent volume



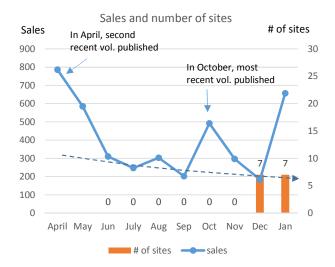
#### (d) Remind effect case 1 a comic for men, the first volume



#### (c) Displacement effect case2 a comic for boys, most recent volume



#### (e) Remind effect case 2 a comic for men, middle volume



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Figure 4(b) and 4(c) are the cases where reach sites listed links one month after the publication. The volumes were published in September, but there were no links to pirated files on the reach site at the month of publication. One month later, ten or eleven reach sites listed the link on the sites, and then the sales declined sharply, deviating from the trend.<sup>7</sup> This deviation — that is, the sharp downshift of sales in October (and November) — is interpreted to be displacement effect.

Figure 4(d) shows the case when pirated files appeared on the reach site long after publication. This comic started in 2012, and this graph represents the sales of the first volume. Since three years passed after the publication, sales were low and gradually decreasing (Note that the left axis is scaled not by log term but number term). A sudden rise of sales in August was caused by the publication of a new volume. The number of reach sites that had the link to this volume was zero until September and began to increase slightly in October and November. The number of reach sites jumped to 11 in December, which is interpreted to have raised the sales in January, deviating from the trend. This deviation from the trend is supposed to be the result of remind effect. Figure 4(e) is a case of a middle volume, the sales of which jumped after piracy files appeared in December. We can interpret this jump as remind effect: consumers read this comic via piracy files and found it interesting, and some of them or others who heard about it through word-of-mouth purchased it.

#### 7. DID regression

Fixed effect cannot cope with endogeneity if the comic's popularity changes during the research periods and site administrators notice it quickly. If a comic suddenly becomes popular in the middle of the research period, both the sales and the number of reach sites (or links to pirated files) will increase simultaneously, which make the number of sites endogenous. Difference-in-difference analysis can cope with this endogeneity problem because it does not use number of sites as an explanatory variable.

We applied the DID model using the massive deletion project, which started in the

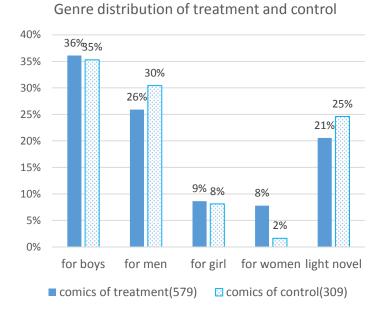
<sup>&</sup>lt;sup>7</sup> It may seems odd that 10 or 11 reach sites suddenly made links to the pirated files. But it is common phenomena. This can happen when there is potential demand to the titles but there is not available pirated PDF file of the title on the Internet. Once someone uploads the pirated file of the title for the first time, it is copied and spread over the many locker sites and many reach sites quickly respond to make links to them. Actually uploaded piracy files at the various locker sites are the exactly same one judging from the files size, picture quality, and page formatting.

middle of research periods as a natural experiment to measure the net effect of deleting pirated files on comic sales. A total of 551 comics were targeted by this deletion, and these made up the treatment group. For control data, we chose 289 comics that were not targets of the massive deletion project.

Control comics were chosen from the title list the of publishers' website. We chose titles which have similar characteristics with treatment comics regarding to the genre, publication year, popularity, length of the volumes. For example we excluded titles with a few number of volumes since most treatment titles consists of five or more volumes.

For difference-in-difference regression, treatment sample and control sample should have the same characteristics except for the exogenous shock. Figure 4 shows the genre distribution of the target comics (treatment) and non-target comics (control). Japanese comics are ordinarily categorized into four genres: for boys, for men, for girls, and for women. Light novel refers to comic-like novels in Japan. As Figure 4 shows, the distribution pattern in terms of genre was similar between the treatment and control groups. Starting year and ending year of the comic were also similar, as shown in the attached sheet in Figure 5. However, sales level was somewhat different; the table shows that total annual sales of the first volume were 5,791 books for the treatment group and 3,220 books for the control group. Publishers were supposed to choose popular titles for this massive deletion project naturally because popular titles are source of their profit. This bias was considered later in the regression of Table 7.

#### Figure 5



	comics of treatment	comics of
	(579)	09)
starting year	2007.6	2007.8
ending year	2014.1	2013.3
total sales of the first volume	5791	3220

Unfortunately, the actual effect of this massive deletion project was limited since deleted files are quickly restored to cyber locker sites, and reach sites rewrite the links. It is often the case that policy intervention against piracy does not affect the level of piracy. (Ushiyama, 2009; Bhattacharjee, Gopal, Lertwachara, & Marsden, 2006). Thus, before estimating the effect of deletion, we had to examine the actual effect of this massive deletion project. According to difference-in-difference approach,

$$Site_{it} = a_i + b_0 * DafterJuly_t + b * Deletion_i * DafterJuly_t + c * Z_{it} + e_{it}$$
(3)

Dependent variable Site<sub>it</sub> represents the number of reach sites of comic i at month t. DafterJuly<sub>t</sub> is equal to one if month t is after July, and Deletion<sub>i</sub> is equal to one if comic i is a target of the massive deletion project. The cross-term coefficient of DafterJuly<sub>t</sub> and Deletion<sub>i</sub> is an effect of the massive deletion project.

Control variables Z<sub>it</sub> are general time trends (time) and nonlinear trends ln(trend) for newly published volumes. Since some cyber locker sites set expiration dates to the files on their server, pirated files tend to disappear gradually even without the deletion request from copyright holders. The time trend variable was introduced to represent this general tendency. Regarding newly published volumes, the number of sites is usually initially zero and increases nonlinearly because uploaded files are copied and spread to other sites quickly. Ln(trend) represents this nonlinear trend.

Table 6 shows the result. As column (2) shows, the cross-term coefficient of completed comics, -0.619, was small but significant. Since the mean of the number of reach sites is 7.163 as shown in Table 3, the decrese ratio is 8.6% (=-0.619/7.163). Therefore, we can say the massive deletion project reduced the number of reach sites successfully for completed comics. Regarding ongoing comics, however, as column (1) shows, the cross-term coefficient was not significant. Ongoing comics are more popular than completed comics because their episodes are released currently in serial comic magazines and often advertised by publishers. Thus, it is natural that pirated files of ongoing comics would be restored to piracy sites more quickly than already completed comics. Quick and frequent re-uploading offset the effect of the massive deletion project. Since the effect of deletion of files is uncertain regarding ongoing comics, we focus on the case of completed comics hereafter.

	(1)	(2)
	Ongoing	Completed
	comics	comics
VARIABLES	Site	Site
Djuly	0.0928	0.466***
	(0.133)	(0.107)
Deletion*Djuly	-0.141	-0.619***
	(0.176)	(0.139)
time	-0.0713***	-0.0166
	(0.0179)	(0.0151)
log(trend)	3.691***	
	(0.249)	
Constant	6.109***	6.924***
	(0.170)	(0.105)
Observations	9,369	4,480
R-squared	0.133	0.009
Number of group	1,345	560

Table 6 Effect of Deletion Project

\*\*\* p<0.01, \*\* p<0.05, \* p<0.1

Table 7 shows the result of difference-in-difference regression. The first two columns represent the basic case. Column (1) is the case of ongoing comics, and column (2) is the case of completed comics. Regarding ongoing comics, the cross-term coefficient of DafterJuly and Deletion was positive (0.703) but not significant. This insignificance corresponds to the insignificance of the deletion effect seen in Table 6. The positive sign suggests that sales might increase with deletion, which is consistent with the result of panel regression seen in section 6.

As for completed comics, the coefficient is significant and -0.083, meaning that the massive deletion project reduced the sale of completed comics by 8.3%. This figure corresponds to the decrease ratio of reach sites, 8.6%. This result implies that pirated files increase the sales of completed comics, which is also consistent with the result of panel regression, confirming the validity of the "remind effect".

	(1)	(2)	(3)	(4)	(5)	(6)
	Ongoing	Completed	Ongoing	Completed	Ongoing	Completed
	comics	comics	comics	comics	comics	comics
	Basic case	Basic case	With monthly	With monthly	100 <sale< td=""><td>100<sale< td=""></sale<></td></sale<>	100 <sale< td=""></sale<>
			dummies	dummies	<5000	<5000
VARIABLES	Log(sale)	Log(sale)	Log(sale)	Log(sale)	Log(sale)	Log(sale)
DafterJuly	0.0837*	0.0480			-0.0547	0.0562
	(0.0439)	(0.0357)			(0.0456)	(0.0405)
Deletion*DafterJuly	0.0703	-0.0830**	0.0672	-0.0800**	0.154***	-0.111**
	(0.0536)	(0.0398)	(0.0537)	(0.0398)	(0.0559)	(0.0469)
time	-0.0584***	-0.0600***	-0.0654***	-0.0653***	-0.0609***	-0.0684***
	(0.00499)	(0.00373)	(0.00630)	(0.00515)	(0.00583)	(0.00544)
log(trend)	-0.722***		-0.724***		-0.410***	
	(0.0569)		(0.0565)		(0.0720)	
newpub1	-0.00776		-0.0294		0.0412	
-	(0.0318)		(0.0325)		(0.0384)	
newpub2	0.112***		0.121***		0.102***	
L	(0.0312)		(0.0318)		(0.0371)	
anime	0.976***	0.862***	0.976***	0.859***	0.864***	0.604***
	(0.0751)	(0.187)	(0.0750)	(0.188)	(0.106)	(0.179)
e promotion	0.508***	0.689***	0.408***	0.656***	0.671***	0.875***
<u> </u>	(0.0362)	(0.0747)	(0.0370)	(0.0762)	(0.0412)	(0.104)
Constant	5.390***	3.446***	5.434***	3.488***	4.689***	4.301***
	(0.0425)	(0.0203)	(0.0610)	(0.0458)	(0.0499)	(0.0278)
Monthly dummies			included	included		
Observations	21,074	10,093	21,074	10,093	12,787	5,493
R-squared	0.166	0.141	0.175	0.146	0.158	0.191
Number of group	2,279	932	2,279	932	1,298	517

Table 7	Difference-in-Difference model

Robust standard errors in parentheses \*\*\* p<0.01, \*\* p<0.05, \* p<0.1

Columns (3)–(6) in Table 7 show a verification of the robustness of this result. Monthly dummies are included in columns (3) and (4) to remove temporary shocks. Columns (5) and (6) limit the data to comics whose sales (measured by annual total sales of the first volume) were between 100 and 5000 in order to consider the difference of total sales level between the treatment and control groups shown in Figure 5. In both cases, cross-term coefficients were positive for ongoing comics and negative for completed comics. The results of the basic case were maintained.

Regarding completed comics, we checked robustness further. We focused on completed comics because the number of sites were significantly decreased by the massive deletion project as show in Table 6 and whether piracy increases legal sales is a controversial topic in copyright issues. As a robustness test, firstly we divided the sample into subgroups and examined the correlation between deletion and sales. If correlation among subgroups was consistent, this suggests that the result was not caused by specific small data, but reflects an overall tendency. This subgrouping check is the approach adopted by Danaher and Smith (2014) to investigate the effects of the shutdown of major cyber locker site, MegaUpload, on DVD sales. Since the penetration of MegaUpload was different among countries, they correlated the countries' sales change after the shutdown with the penetration ratio of MegaUpload and found a positive correlation. In other words, they divided the worldwide consumers into country subgroups and checked the correlation between the degree of intervention (penetration of MegaUpload) and outcome (DVD sales) in the difference-in-difference model. We applied this approach as well.

Firstly, we divided the completed comics by ending year into five groups and ran DID regressions of sites and sales. We then checked correlation of cross-term coefficients. Let b<sub>site</sub><sup>y</sup> be coefficients of cross term, DafterJuly\*Deletion, of the regression of the reach site (model (2) of Table 6) and let b<sub>log(sale)</sub><sup>y</sup> be cross-term coefficients of the regression of the sales (model(2) of Table 8) when the data is limited to comics ending in year y. Both terms measure changes caused by the massive deletion project, and b<sub>site</sub><sup>y</sup> represents the change in number of sites of comic titles ending in year y. If piracy reduces legal sales, b<sub>site</sub><sup>y</sup> and b<sub>log(sale)</sub><sup>y</sup> have a negative correlation in terms of y because, when more sites are reduced for the comic group y, then more sales increase for the same comic group y. To the contrary, if piracy increases legal sales, they correlate positively.

Figure 6(a) shows the result. The horizontal line is  $b_{site}y$ , change in number of sites, and the vertical line is  $b_{log(sale)}y$ , change in sales. In our findings, we observed weak positive correlation, which means that comics deleted more are sold less in terms of ending year grouping. This result suggests that piracy increases legal sales in case of completed comics.

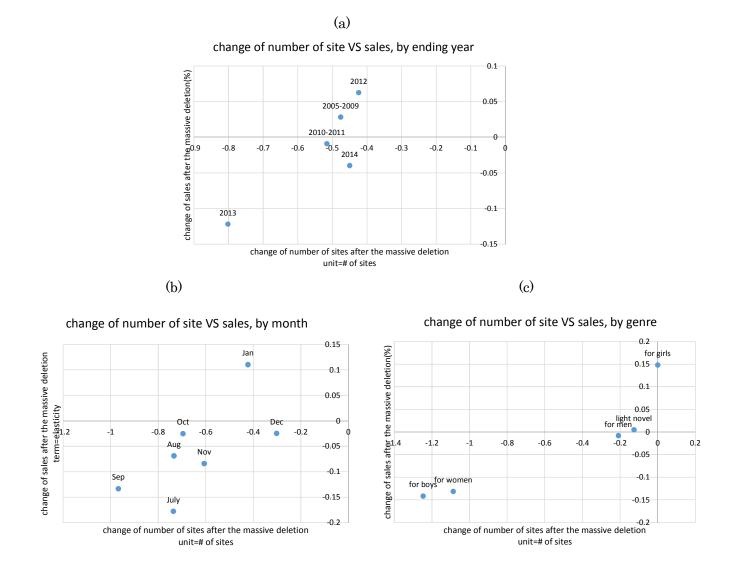


Figure 6 Robustness check of remind effect of completed comics

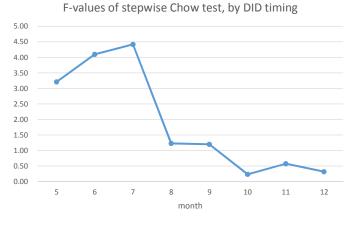
Secondly, we broke down the cross-term coefficient into monthly terms by replacing DafterJuly<sub>t</sub> with Djuly<sub>t</sub>, Daug<sub>t</sub>, Dsep<sub>t</sub>, ..., Ddec<sub>t</sub>, which are equal to one if t is equal to the current month. Since the performance of the massive deletion project may change month to month, the degree of reduction of reach sites also may change over time. If the sales of comics positively correlate with the change in site reduction by monthly terms, this is supporting evidence of the remind effect hypthothesis. The results are shown Figure 6(b), indicating positive correlation, which means that sales decreased more in the month when reach sites decreased more. This result again suggests piracy increases the legal sale of completed comics.

Thirdly, we let the coefficients change depending on comic genres. As previously

mentioned, the Japanese comic industry has five genres: for boys, for men, for girls, for women, and light novels. Genre dummies Dg were introduced to correspond to these genres and effect of the mass deletion to sites and sales was measured separately by genres. Vertical axix of Figure 6(c) shows coefficients of effect on sales and horizontal axis shows effect on site by five genres. Correlation of these coefficients are positive again. Therefore, if pirated files of comics of a genre are deleted more than other genres, the sales of comics of the genre decrease more than other genres. This result suggests that piracy increases legal sales with regard to completed comics.

In Figure 6(a)(b)(c), we broke down cross-term coefficients of difference-in-difference regression by subgrouping data with the purpose of examining the consistency of degree of policy intervention (deletion) and its outcome (sales). All graphs show that correlation is positive; thus, they are consistent with the remind effect hypothesis. No matter which subgrouping is used, either by ending year, by timing, or by genre, more deletion leads to less sales, which means piracy stimulates sales as far as completed comics are concerned.

As second robustness check I will conduct placebo test. Although the placebo test with respect to timing of inervention is common for difference-in-difference regression, we could not apply it here because the sample size in the time series was too small (12 months). As for timing we can only apply ordinary stepwise Chow test to see whether July was the proper timing for applying DID regression. The null hypothesis was that coefficients of both DafterJuly and DafterJuly\*Deletion were zero in the regression (2) of Table 7. Figure 7 shows the F-value of Chow test changing the hypothetical starting point of massive deletion from May to December. Clearly, July is the most probable point at which structural change occurred, which suggests the July was the proper point to apply difference-in-difference regression.

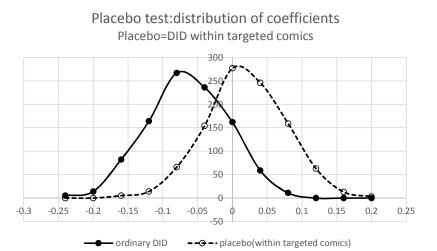


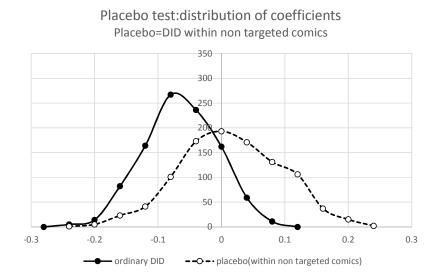


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Instead of placebo test on timing, we applied placebo test by randomly choosing the treatment and control samples. We randomly drew one third and another one third from the same targeted comics and assign the one as a treatment and the other as a control hypothetically, applying difference-in-difference regression. This draw was conducted a thousand of times and a thousand of draws provided us distribution of estimated coefficients of cross term of the placebo difference-in-difference regression, which is shown as the dashed line in Figure 8(a). Solid line is the the case of the oridinary DID regression, where treatment is randomly chosen one third of targeted comics and control is one third of the nontargted comics. Two distributions are clearly defferent suggesting that the negative coefficients of pracebo TID is applied within non targeted comics, indicating also that coefficients of pracebo tests show that the negative coefficient of the ordinary DID regression is not accidental result from those of the ordinary DID regression. These placebo tests show that the negative coefficient of the ordinary DID regression is not accidental result to the ordinary DID regression is not accidental result but robust result not depending on the sample selection







#### 8. Summary and Discussion

In this paper, we examined the effect of piracy on the comic book market in Japan using direct measurement of product level piracy ratio and a massive deletion project as a natural experiment. We found the piracy effect was heterogeneous depending on whether the comic series was ongoing or completed. Piracy decreased the legitimate sales of ongoing comics but stimulated legitimate sales of completed comics: displacement effect was dominant for ongoing content, and advertisement effect was dominant for completed content. We can interpret this result as an indication that the piracy reminds consumers of past comics that are no longer promoted by publishers.

An interesting question is whether the net effect of piracy is negative or positive to total legitimate sales. Unfortunately, we were unable to calculate this because the sales share of ongoing comics and completed comics is unknown. The loss of displacement effect is far larger than the gain of remind effect per individual book base because ongoing comics usually sell over ten thousand copies in the newly published month while most completed comics series sell only several hundred in a month. However, number of volumes of completed comics outnumbers newly published volumes of ongoing series. Which effect is dominant depends on the total sales share of ongoing and completed comics, and this figure is not available.

For the purpose of discussion, as a back-of-envelop estimation, I tried brief calculation of total effect of piracy on the assumption that our list of comic titles were representative of total comics. If our list of comic title is representative of whole comic titles, we can estimate the total effect of piracy by simply appling the estimated elasticities on the observed total sales of our sample. The result showed that negative effect is larger than positive effect, indicating that piracy reduces the total revenue of publishers (see Appendix).

An important point we should stress here is the policy implication of the heterogeneity of piracy effect. If the effect of piracy is heterogeneous, it is not the best solution to shut down the piracy sites uniformly but to delete harmful piracy files selectively if possible. In this case, deleting piracy files of ongoing comics only is the first best strategy for publishers regardless of whether the total effect is positive or negative, because the availability of piracy files of completed comics is beneficial to both publishers and consumers.

Heterogeneity is not limited to the book industry. In 2007, Blackburn showed that file sharing decreased CD sales of big artists whereas it increased CD sales of unknown minor artists. Bhattacharjee et al. (2007) reported that file sharing reduced survival time in the Billboard CD album ranking for low-ranked albums but did not hurt topranked albums. Also Peukert et al. (2017) showed that the shutdown of Megaupload had a differential effect on box office revenues of wide-release vs. niche movies. Mixed results in the literature of piracy studies so far may be caused by the heterogeneity of products. Product level research that takes this heterogeneity into consideration may shed some light on the effects of piracy. This paper's contribution is the addition of new dimension to the heterogeneity — present content and past content.

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#### Appendix : Total effect of piracy

I will use the elasticities of piracy sites on sales with regard to four comic categories in Table 4, which are shown in the column (a) in Table A1. Column (b) is number of volumes of the category and (c) shows the total sales of the volumes.

We need adjustment term because the comic consists of multiple volumes. Average number of volumes is 15.7. Regardgin to ongoing comics, we assume the elasticity of recent two volumes are applied to recent two volume only and other volumes' elasticity is equal to that of 1st and middle volumes.<sup>8</sup> With regard to completed comics, the elasticity of the first and middle volumes is applied to the first half of the volumes, and the elasiticity of recent two volumes is applied the last half of the volumes.<sup>9</sup> Column (e) is adjusted total sales of the categories by multiplying adjustment term (d) with total sales (c).

<sup>8 6.850=(15.7-2)/2</sup> 

<sup>9 3.925=15.7/2/2</sup> 

Column (f) shows the effect of 1% increase of number of piracy sites by multiplying elasticity (a) with adjusted sales (e). Adding the column (f) provides with total effect of piracy, which is clearly negative. One percent increase of number of piracy site decreases the total sales by 27,196 volumes. If the piracy sites become double, like 5 to 10 which is often the case with, the100% increase of number of piracy sites decrease the sales by 271,600 which is approximately 3.1% of the total sales (0.031=271600/86377305)

		(a)	(b)	(c)	(d)	(e)	(f)
		coefficients	n	total sales	adjustment	=(c)*(d)	=(e)*(a)/100
ongoing	recent 2 volumes	-0.0967	1114	26,892,212	1.000	26,892,212	-26,005
	1st and middle	-0.0397	1169	7,244,668	6.850	49,625,976	-19,702
completed,	recent 2 volumes	-0.0401	552	1,493,701	3.925	5,862,776	-2,351
	1st and middle	0.522	557	1,018,176	3.925	3,996,341	20,861
						86,377,305	-27,196

Table A1 Back-of-Envelop Calculation of the Total Effect of Piracy

Critical assumption of this calculation is that the list of comic titles in our survey is representative of the whole current comic titles. This assumption is probably misleading because the number of title in column (b) in Table A1 suggests that completed comics is under-sampled. Column (b) shows number of volume of completed comics is five hundreds which is a half of the ongoing comics, but number of completed comics should be equal or outnumber the ongoing comics because completed comics are accumulated sum of past comics, even if some of them become out of print or disappear from the piracy site because of out-of-dated fashion. If the true number of volumes of completed comics is double, the total effect of the piracy is not clear since the total effect of the colum (f) is nearly zero. Since we don't have enuogh information of the ratio of ongoing and completed comics, this result is tentative.