Supply responses to digital distribution: Recorded music and live performances

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1. Introduction

New information technologies in the past decade have radically changed the methods of distributing information goods (i.e., products that can be digitalized). These technologies make new distribution channels available to consumers, but also raise the risk of illegitimate redistribution. Understanding how firms can create and distribute information goods while still protecting their intellectual property has been the core issue of many policy debates, including those surrounding the passage of the Digital Millenium Copyright Act (DMCA), the Uniform Computer Information Transactions Act (UCITA), and a landmark US Supreme Court case (MGM v. Grokster). The fundamental economic concern is that redistribution technologies may threaten markets for information goods by making it difficult for producers to capture the returns to their investments.

However, concerns about the viability of markets for digitally redistributable products may be tempered if firms can recover their investments through the sale of complementary, non-digital goods. Redistribution of the digital good may increase demand for the complementary good, partially offsetting the losses due to illegal redistribution of the digital good. The implication, as argued by Teece (1986), is that public policy aimed at promoting innovation should not ignore the impact of an innovation on goods or assets that are complementary to it.

In this paper we study firms’ responses to digital redistribution technologies in the specific context of the music industry. Large-scale file-sharing of recorded music began with the entry of Napster in 1999, and recorded music has been at the forefront of debates about the impact of digital distribution ever since. Several empirical studies have focused on measuring the extent to which illegal downloads displaced legal sales. In contrast, our primary
focus is to examine how file-sharing affected the complementary market for live concert performances. To address this question, we construct an extensive dataset of concert events and album sales. The concert data contain information on over 200,000 concerts between 1995 and 2004 performed by over 12,000 artists. We are able to combine detailed sales data with the concert data for a sample of 1806 artists.2

We expect two kinds of effects of recorded music file-sharing on live concert performances. One effect is a demand shift: if recorded music and live performances are complements (e.g., because a concert is more enjoyable if you have listened to the recorded music ahead of time), then increases in the consumption of recorded music due to file-sharing should lead to increased demand for live performances. The second effect is a supply shift: to the extent that file-sharing reduces the profits from selling recorded music, we would expect artists to reallocate effort toward concert tours and away from recording new albums. Both of these effects imply that concert activity should increase after the entry of Napster in 1999 and the adoption of large-scale file-sharing. In fact, the number of concert events increased sharply beginning in 2001, as shown in Fig. 1. The surge in the number of concerts is a sharp contrast to the concomitant decline in album sales, which the record industry blamed on file-sharing.

Our objective is to document various trends in the production of recorded music and live performances, and ask whether those trends are consistent with the two effects described above. At the most basic level, we look for breaks in trend like the one shown in Fig. 1. Further, we examine variation in trends across artist types. Most notably, we show that concert revenues of smaller, more obscure bands increased by more than for larger bands after the onset of file-sharing. Smaller bands also experienced a smaller decline in album sales. This suggests that broad redistribution may have increased awareness of obscure bands by making their music available from more sources and at a much lower cost (or for free in the case of illegal file-sharing). Broader awareness of these bands would increase demand for their concerts. Similarly, while file-sharing may displace some album sales, for small bands this effect may be mitigated by increased sales resulting from a larger fan base, again due to increased awareness. Conversely, file-sharing may have a small impact on concert demand for large, “superstar” bands who were already well-known and whose music was already widely played. For large bands we expect file-sharing to substantially displace album sales with little mitigating expansion in overall listenership.

We examine this hypothesis by ranking artists by concert revenues and album sales, and then calculating growth rates in concert revenues and album sales by rank (i.e., we calculate the growth rate by comparing the concert revenues of the 5th-ranked artist in 1996 to the 5th-ranked artist in 1995). Consistent with prior studies, we find a substantial decline in album sales following the entry of Napster. We also find that concert revenues grew at a substantially higher rate following the entry of Napster. Furthermore, we find that concert revenues for the highest-ranked artists were mostly unaffected by the entry of Napster, while revenues for smaller, lower-ranked artists grew at a much faster rate following Napster than before Napster. On the other hand, album sales for top-ranked artists fell dramatically following the entry of Napster, while album sales for small artists experienced only a slight decline. Hence, superstar artists appear to receive little or no benefit from file-sharing in terms of increasing demand for their concerts, presumably because people were widely aware of their music prior to file-sharing. For smaller

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2 We collect data in both markets going back to 1993. In several of our analyses, we construct variables (e.g., growth rates) that limit our analyses to start no earlier than 1995, and for consistency we report all analyses for the period of 1995 and later. The general implications of the analyses are not sensitive to the inclusion or exclusion of data prior to 1995.
artists, however, our findings are consistent with file-sharing increasing awareness of the artists’ music and consequently increasing demand for live concert performances by those artists, as well as mitigating any loss in recorded music sales from illegal downloading.

While our study focuses on the music industry, the economic phenomena we analyze are clearly relevant in many other markets. For example, digital copies of movies may cut into home video sales, but may also lead to higher demand for movie-related merchandise. An author's royalties from book sales may be reduced if the book is digitally shared, but the increased readership may lead to profits on the lecture circuit. Mass sharing of a pirated software program may displace paid licenses for that program, but may also generate increased sales of complementary physical products or technical support services.

The paper proceeds as follows. In the next section we provide a brief description of the music industry, and we describe our data sources. In Section 3 we describe aggregate trends in concert activity and recorded music sales; in Section 4 we analyze more detailed artist-level data for the sample of 1806 artists for whom we observe both concert revenues and CD sales. Section 4.2 presents robustness checks, and Section 5 provides a discussion of our findings and concludes.

2. Background and data

2.1. Music industry background

Professional music industry background

Recorded music is produced under contract with a record label: the artist records an album as a work-for-hire, and the record label markets and distributes the album. Typical production costs during our sample period were in the neighborhood of $100,000–$250,000, and industry executives report that marketing and distribution costs often eclipsed the cost of production. The standard contract is a royalty contract: the artist is paid royalties on album sales, and receives an advance against those royalties in order to cover living expenses and studio costs during the production of the album. Royalty rates range between 10% and 18% of retail, with the typical rate being 12%; however, artists earn somewhat less than this due to various deductions that are usually built into the contract. For the time period we study, a reasonable estimate is that the artist earns around $1.00 for every CD she sells.4 Record labels during this period hold a negligible stake in the live performance business. Although labels usually offered some nominal touring support to new artists as part of the recording contract, and sometimes coordinated with concert promoters to advertise a show, they do not take a share of the touring revenues.5 More recently, these contracts have begun to change, so that recording labels contract explicitly over some share of the artist's touring receipts. Artists' live performances are coordinated and underwritten by concert promoters. The promoter finances almost every aspect of the concert production, including renting the venue, paying the artist and staff, and advertising. Artists are paid as a percentage of ticket revenues, subject to some minimum (called the "guarantee"). Artists also make money from merchandise sales; for some artists this can be a significant component of the net earnings. A typical deal gives 70–80% of merchandise revenues and 70–85% of the gross ticket revenues to the artist, although the actual percentages may be somewhat lower because various deductions are made to the gross ticket revenues before the artist's cut is taken.

Although artists have virtually no say in the pricing of recorded music, most industry sources identify the artist as the primary agent with responsibility for setting concert ticket prices. The artist and/or artist's manager sets prices in consultation with the promoter and venue owner. The parties can have conflicting incentives; for example, aside from the rental fee for the venue, the venue owner's revenue comes primarily from concessions and parking, so they tend to push for low ticket prices in order to fill the house.

In May of 1999, the software program Napster introduced an easy-to-use interface by which consumers could share and download digital copies of songs. Napster and similar programs represented a dramatic shift in the distribution technology for recorded music.6 Napster gained currency quickly, with a reported user base of over 20 million unique accounts at its peak and over a half million unique IP addresses connected at any given time on a routine basis.7 The Recording Industry Association of America (RIAA) claimed that the presence of Napster eroded sales of CDs by facilitating copyright violations, and sued to have Napster dismantled in December of 1999. In 2003, the RIAA began suing individual participants of file sharing networks, and subsequent activity on these networks was reported to have declined.8

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3 Some very successful songwriters also earn significant revenues from music publishing fees, and some star artists have substantial income from endorsements, but the typical artist relies mostly on recorded music sales and concerts.

4 Instead of using a standard royalty contract, some artists negotiate “penny contracts” specifying artist payments as a fixed dollar amount per CD sold. The typical artist share in these contracts is reportedly $1.25 per CD; however, artists who negotiate such contracts have more bargaining power than the average artist.

5 Tour support is typically a recoupable expense, but it is recouped from recorded music revenues. This convention may be a holdover from an earlier era: historically, labels subsidized concert tours only as a way of promoting albums, and concerts were often not expected to be profitable on their own.

6 Although earlier technologies also allowed for illegitimate reproduction (e.g., cassette tapes are easily copied), they were much more limited in scope, and typically had greater quality degradation.

7 Original source: CNNMoney 2000. For an excellent review of the industry and the timing of filesharing events specifically, see Blackburn (2004).

8 The Supreme Court ruling in MGM v. Grokster in June 2005 represented a significant legal victory for the RIAA, as the court held that distributors of file-sharing software could be held secondarily liable for copyright infringements facilitated by their software, essentially allowing the RIAA to go beyond merely suing individuals who shared files illegally to suing the companies whose software enables the sharing. In addition to the legal front, the music industry has also battled file-sharing on the technological front, using various encryption and digital rights management technologies to curb the flow of illegal music downloads. Park and Scotchmer (2004) analyze the impact of such technologies on the pricing of digital goods. Legal channels of digital music distribution are, by now, becoming well established. Most notably, iTunes launched in October, 2003.
2.2. Data

The data we use in this study come from several sources. The data on concerts come from Pollstar, a company that tracks virtually all concert activity in the United States. The data describe 227,230 concert events performed by 12,356 artists in the period 1993–2004. For concerts performed between 1993–2002, the data provide detailed box office information, including tickets sold, total ticket revenues, and high and low ticket prices. For 2003 and 2004, we observe the dates, locations (city and venue name), and identities of all performing bands, but we do not have data on box office receipts. Although data for years past 2004 are obtainable, we truncate at 2004 because the company implemented a significant change to its reporting mechanism in 2005 that makes later years’ data difficult to compare with the period in which we are primarily interested (i.e., the years around 1999).

The dramatic increase in concert activity shown in Fig. 1 is based on Pollstar data. Given the change in reporting in 2005 (and consequent change in coverage), one might worry that the increase in concert events after 2000 also reflects a change in data coverage rather than a change in actual concert activity. To address this concern, we conducted an extensive audit of the Pollstar data during the years of 1996–2002 for concerts in the Boston area using a local weekly newspaper that has a reputation as the best guide for music in the area. Pollstar did not have complete coverage of all events (particularly events at small venues). The coverage rate increased slightly over time, but the increase was gradual. Most importantly, there was no evidence that the coverage rate increased sharply after 2000.9

Our data on album sales come from Nielsen SoundScan, a company that tracks music sales at the point of sale.10 Some of the results we report below are based on aggregate sales by designated market area (DMA), covering the years 1993–2002.11 We also observe highly disaggregated data for a subsample of 1,806 artists who we can match to the concert data from Pollstar. We refer to these artists as our “matched sample.” For these artists we observe weekly CD sales by DMA at the individual album level; for the analyses in this paper, we summarize over the artists’ albums and simply examine total album sales by artist/DMA/week. The matched sample may not be perfectly representative of the broader universe of artists, and we discuss sample selection issues in Section 4 along with running concert analyses both on the matched sample and the full Pollstar data.

We collect additional data on recorded music for individual albums from MusicBrainz, an online database that tracks detailed album information for official album releases for the purpose of documenting or “tagging” downloaded music tracks.12 The MusicBrainz database has extensive coverage of recorded music releases, tracking 240,000 albums from 100,000 artists, and provides information on the date of release, length and title of each song, album credits, and so on. The data from MusicBrainz are useful for documenting the timing of album releases, and also for characterizing the population of recorded releases over time.13

In addition to the detailed data on concerts and recorded music, we supplement our analyses with a number of characteristics about artists and recorded releases that are collected from several other sources. We use artist characteristics from allmusic.com’s online database of artists, and data from Recorded Industry Association of America (RIAA) to measure cumulative album sales prior to 1993 for artists that were established before that date. We also use data from BigChampagne (collected during 2007) to measure a cross-section of downloading activity across artists and cities.14

3. Aggregate trends

Since file-sharing technologies made millions of songs freely downloadable over the internet, they were naturally expected to displace legal sales. Most empirical studies have found evidence of this displacement. However, while file-sharing decreased legal sales, it almost certainly increased the overall consumption of recorded music. Evidence from time-use surveys indicates a dramatic increase in music listening between 1998 and 2001. In one survey, respondents in 2001 reported spending three times as much time listening to music as respondents from 1998. More tellingly, among respondents who reported having below-median internet usage rates, the increase in music listening was negligible (just over 10%), whereas the increase for those with above-median internet usage was more than tenfold.15 Since recorded music and live performances are complementary, an increase in the consumption of recorded music should increase the demand for concerts: the more people who are listening to an artist’s music, the larger is the market for a concert by that artist.

In addition to this demand-side effect, file-sharing may also have shifted artists’ incentives on the supply side. To the extent that file-sharing eroded the profitability of selling recorded music, artists had an incentive to reallocate effort away from recording new albums, instead performing more frequent and/or more extensive concert tours.

As a first step in documenting the empirical relevance of these effects, in this section we describe aggregate trends in concert activity and music sales in the years before and after file-sharing technologies became pervasive. Time trends alone cannot establish any causal link; our purpose in this section is simply to examine whether the

9 The local weekly newspaper used for the audit was The Boston Phoenix. Details of the audit are available from the authors upon request.
10 For our sample period, SoundScan’s coverage rate was reported to be approximately 80% of all music sales.
11 A DMA is similar to an MSA.
12 The MusicBrainz database is used by a wide range of complementary “tagging” software programs, and is widely cited in the music community as a reliable source of information on recorded songs and albums.
13 This allows for a comparison of our matched sample to the population of releases over time.
14 We cite these additional sources and provide additional detail throughout the analyses as appropriate.
15 Based on internet-accessible data from National Time Diary Studies conducted by the Survey Research Center at the University of Maryland: see http://www.popcenter.umd.edu/sdaweb/diary9801/Doc/Diar.htm.
Table 1: Concerts: changes over time.

<table>
<thead>
<tr>
<th>Year</th>
<th>Number of concerts</th>
<th>Number of artists on tour</th>
<th>Concerts per artist</th>
<th>Cities per artist</th>
<th>% In largest 20 DMAs</th>
<th>Tickets per concert</th>
<th>Average price</th>
</tr>
</thead>
<tbody>
<tr>
<td>1995</td>
<td>16,027</td>
<td>2159</td>
<td>7.42</td>
<td>15.66</td>
<td>58.86</td>
<td>2841.03</td>
<td>26.32</td>
</tr>
<tr>
<td>1996</td>
<td>17,222</td>
<td>2309</td>
<td>7.46</td>
<td>15.88</td>
<td>57.37</td>
<td>2872.94</td>
<td>25.38</td>
</tr>
<tr>
<td>1997</td>
<td>16,971</td>
<td>2348</td>
<td>7.23</td>
<td>15.38</td>
<td>56.78</td>
<td>2923.35</td>
<td>27.84</td>
</tr>
<tr>
<td>1998</td>
<td>15,503</td>
<td>2361</td>
<td>6.57</td>
<td>14.49</td>
<td>58.08</td>
<td>3128.90</td>
<td>29.31</td>
</tr>
<tr>
<td>1999</td>
<td>15,077</td>
<td>2297</td>
<td>6.56</td>
<td>13.99</td>
<td>60.53</td>
<td>3246.29</td>
<td>33.24</td>
</tr>
<tr>
<td>2000</td>
<td>15,065</td>
<td>2309</td>
<td>6.52</td>
<td>14.35</td>
<td>61.18</td>
<td>3293.64</td>
<td>35.52</td>
</tr>
<tr>
<td>2001</td>
<td>19,425</td>
<td>2894</td>
<td>6.71</td>
<td>13.85</td>
<td>60.91</td>
<td>2708.98</td>
<td>35.26</td>
</tr>
<tr>
<td>2002</td>
<td>22,033</td>
<td>3292</td>
<td>6.69</td>
<td>14.25</td>
<td>64.03</td>
<td>2459.34</td>
<td>36.02</td>
</tr>
<tr>
<td>2003</td>
<td>20,791</td>
<td>3683</td>
<td>5.65</td>
<td>14.28</td>
<td>59.99</td>
<td>2459.34</td>
<td>36.02</td>
</tr>
<tr>
<td>2004</td>
<td>24,103</td>
<td>4212</td>
<td>5.72</td>
<td>14.68</td>
<td>58.37</td>
<td>2459.34</td>
<td>36.02</td>
</tr>
</tbody>
</table>

Based on Pollstar data. Cities per artist is the average number of different major markets (DMAs) artists performed in, conditional on performing at least one concert. Prices in the last column are deflated to 1999 dollars using the CPI.

patterns of change within the music industry are consistent with the anticipated effects of file-sharing.

3.1. Live performances

Fig. 1 documents the sharp increase in the number of live performances after the year 2000. Table 1 provides more detail on trends in live performances between 1995–2004.\(^\text{16}\) The first column reports the number of concerts, and matches Fig. 1. The second column reports the number of artists on tour, which also increases sharply after 2000. The number of concerts per artist is reported in the third column. This number is falling before 2000, and then rising in 2001–2002 and falling again in 2003–2004. We suspect that the entry of many new artists, especially in 2003–2004, means that the marginal artist on tour performs fewer concerts in those years. The number of cities in which an artist tours falls from 1995–1999, and then levels off. The percentage of concerts performed in the 20 largest cities in the US is relatively stable over time. The number of tickets per concert rises from 1995–2000, and then falls. Combined with the first column, this suggests a trend toward a larger number of smaller concerts. Note that the fall in the number of tickets per concert is smaller than the rise in the number of concerts performed. Thus, total ticket sales increased over the period.

The table also shows that average ticket prices rose steadily (especially after 1999). A likely explanation for the simultaneous increase in both price and quantity is that demand for concerts increased; a supply-shift alone would not result in both higher price and quantity unless other related expenses changed (e.g., the cost of transportation fell significantly for concert-goers).

In addition to changes over time, we also observe cross-sectional variation in concert performances over time across cities. The top panel of Table 2 reports average annual growth rates in concert performances for the 1996–1998 and 2000–2002 periods for different types of cities over time. We classify cities based on two different measures of the likely importance of file-sharing. In the first classification, we designate a DMA as “low (high) broadband” if its broadband penetration is below (above) the median. Second, we classify cities into high- and low-download markets based on whether a city’s downloading activity was disproportionately high or low relative to population, based on the BigChampagne data from 2007.\(^\text{17}\)

Growth in concert performances accelerated sharply for both high- and low-broadband markets in the post-Napster period, and the acceleration was significantly more pronounced in the high-broadband markets. A similar pattern holds for the high- vs. low-download classification. While we do not know how accurately these groupings proxy for actual file-sharing, the patterns in the table are at least consistent with our hypothesis that increases in concert activity were driven partly by the arrival of file-sharing.

3.2. Recorded music

The bottom panel of Table 2 shows analogous comparisons for growth rates in album sales. Annual sales were growing at double-digit rates in the years just before file-sharing, but began shrinking after file-sharing. In contrast to concert performances, however, the trends were essentially the same in high- and low-broadband markets (and high- and low-download markets). Thus, sales declines were not more pronounced in the markets that one might associate with greater file-sharing activity. This could reflect the coarseness of our proxies for file-sharing and the difficulty of measuring the degree of sales displacement that results from file-sharing. (The studies that have found convincing evidence of a displacement effect have used individual-level data, not market-level comparisons.) Regardless of its cause, the decline in album sales after 1999 may have reduced artists’ incentives to produce recorded music, and the apparent increase in demand for concerts would have further pushed artists to reallocate effort away from recording albums toward performing concerts. One of the empirical implications of this realloca-

\(^{16}\) As noted earlier, we focus on the time periods 1995–2002 or 1995–2004 depending on availability of the relevant variables. Including data for earlier years (e.g., 1993 and 1994) does not impact the general implications of the analyses and is not feasible for the analyses in Section 4.

\(^{17}\) Specifically, we regress the log of total downloads (by city) on population rank, and then classify cities based on the residuals. (Cities with residuals above the median are the “high-download” cities.)
sharp breaks in trend. Both series increase over time; neither exhibits any
new albums over time, again based on the MusicBrainz data-
CDs.18 Table 4 documents the entry of new artists and
have been able to profitably exist, and also to release
recorded music to potential listeners, more artists may
come more profitable because of the greater accessibility of
and release fewer albums. On the other hand, if touring be-
tent that file-sharing reduced the profitability of recorded
music, we could expect record labels to sign fewer artists
ing artists, file-sharing may also have affected the entry of
well.

the 1995–1999 period as
is no obvious break in trend around 1999–2000. The trend
increased after 1999, primarily because more artists took
(tme until next release). The lag between album releases
patterns are similar if we use a forward-looking measure
based on the MusicBrainz database. The table reports a
Table 3 reports trends in the time between album releases,
Table 3 examines the album production of exist-
ing artists, file-sharing may also have affected the entry of
new artists. This effect is more subtle, however. To the ex-
tent that file-sharing reduced the profitability of recorded
music, we could expect record labels to sign fewer artists
and release fewer albums. On the other hand, if touring be-
came more profitable because of the greater accessibility of
recorded music to potential listeners, more artists may
have been able to profitably exist, and also to release
CDs.18 Table 4 documents the entry of new artists and
new albums over time, again based on the MusicBrainz data-
base. Both series increase over time; neither exhibits any
sharp breaks in trend.

18 Relatedly, if digital technology lowers the cost of producing recorded
music, we may also get greater entry into the market for recorded goods in
spite of the potential for file-sharing to have some displacement effect.

4. Heterogeneity across artists

The impact of file-sharing on concerts and album sales
may differ substantially between large and small artists. File-sharing may increase awareness of smaller, more ob-
scure artists and their music by making the music available
from more sources and at a much lower cost (or for free in
the case of illegal file-sharing). Broader awareness of these
artists should be reflected in increased demand for their
concert performances. Similarly, while file-sharing may
offset some album sales for small artists this may be miti-
gated in part by increased sales from the larger potential
fan base that may result from increased awareness of those
artists. On the other hand, file-sharing may have a rela-
tively small impact on the awareness of music for large,
“superstar” artists. The music for these artists was already
widely played prior to file-sharing, and thus file-sharing
may have had little impact on concert demand for these
top performers. Similarly, file-sharing is likely to primar-
ily displace album sales for large bands with little or no com-
pensating increase from a potentially larger fan base. In
summary, we expect small bands to experience a greater
increase in concert revenues and less of a decline in album
sales than large, highly popular bands following the entry
of Napster.

Analyzing artist-level data is challenging because ar-
tists’ album sales and concert revenues are extremely vol-
tile over time, making it difficult to appropriately define
the unit of observation for empirical analysis. As with
many other media goods (e.g., books, movies, video
games), demand for recorded music tends to peak at or
near the release date and then decay rapidly over time.
Furthermore, production occurs in discrete jumps—artists
only release new albums every 1–3 years, and may only
go on tour every 1–4 years—so analyzing data at the art-
ist-year level is problematic.

For the purposes of comparing the effects of file-sharing
on large vs. small artists, our solution to this problem is to
define the unit of observation as a “year-rank” pair. That is,
rather than analyze the sales of a given artist from one year
to the next, we analyze the sales of the nth-largest artist
from one year to the next (where the nth-largest artist in
year t may be different from the nth-largest artist in year
t + 1). For example, to track concert revenues over time
we first rank artists by their total concert revenues in each
year. We then analyzes changes in revenue at each rank
over time. In essence, we are asking whether the changes

<table>
<thead>
<tr>
<th>Year</th>
<th>Album sales</th>
</tr>
</thead>
<tbody>
<tr>
<td>1995</td>
<td>59.60</td>
</tr>
<tr>
<td>1996</td>
<td>56.39</td>
</tr>
<tr>
<td>1997</td>
<td>56.38</td>
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<tr>
<td>1998</td>
<td>54.38</td>
</tr>
<tr>
<td>1999</td>
<td>55.36</td>
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<tr>
<td>2000</td>
<td>54.39</td>
</tr>
<tr>
<td>2001</td>
<td>53.23</td>
</tr>
<tr>
<td>2002</td>
<td>51.38</td>
</tr>
<tr>
<td>2003</td>
<td>50.75</td>
</tr>
<tr>
<td>2004</td>
<td>50.09</td>
</tr>
</tbody>
</table>

Table 2
Average annual growth: concert performances and album sales.

<table>
<thead>
<tr>
<th>Year</th>
<th>Concert performances</th>
<th>Album sales</th>
</tr>
</thead>
<tbody>
<tr>
<td>1996–1998</td>
<td>4.4</td>
<td>20.2</td>
</tr>
<tr>
<td>2000–2002</td>
<td>36.3</td>
<td>19.0</td>
</tr>
</tbody>
</table>

Cells report averages (across DMAs) of the percentage growth rate for the designated time period. Low vs. high broadband distinction is based on Forrester Research broadband penetration measure (percent of households with broadband internet in 1999). Low vs. high download distinction is based on BigChampagne data; see text for explanation.

Table 3
Time between album releases.

<table>
<thead>
<tr>
<th>Album released in:</th>
<th>Years since last release</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>0–1</td>
</tr>
<tr>
<td>1995</td>
<td>59.60</td>
</tr>
<tr>
<td>1996</td>
<td>56.39</td>
</tr>
<tr>
<td>1997</td>
<td>56.38</td>
</tr>
<tr>
<td>1998</td>
<td>54.38</td>
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<tr>
<td>1999</td>
<td>55.36</td>
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<td>2000</td>
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<td>2003</td>
<td>50.75</td>
</tr>
<tr>
<td>2004</td>
<td>50.09</td>
</tr>
</tbody>
</table>

Times between releases calculated from MusicBrainz database. Cells report the percentage of albums that fall in each category in each pair of year based on the number of calendar years elapsed since the artist’s previous album release. (Debut albums are excluded.)
in concert revenues or album sales differed across the distribution of artist size.

To analyze the impact of file-sharing on a consistent basis for both concerts and album sales, we create a matched sample for the 1806 artists for whom we have detailed album sales and concert revenue information. All concert-based analyses are carried out for the same set of artists used to analyze album sales. We also implement the concert-based analyses on the full Pollstar concert data to ensure that any findings based on the matched sample are also consistent with the population of concerts more generally. The concert revenue findings based on the matched sample and the full Pollstar data have very similar implications, suggesting that our findings for the matched sample are representative of all artists more broadly.

Our main outcome variables of interest are growth rates for concert revenue and album sales, and we analyze these outcomes separately at two levels: (1) the national level and (2) groups of DMAs based on broadband internet penetration in 1999.19 At the national level, we calculate growth rates for concert revenues by aggregating annual concert revenues to the national level and ranking each artist based on his aggregate concert revenues in that year. Using these individual year-rank observations, we calculate a growth rate as the difference between the log of concert revenues in the current and preceding years for artists of the same rank. We repeat the exercise using album sales.

Table 4
Recorded music: changes over time.

<table>
<thead>
<tr>
<th>Year</th>
<th>Unit sales (millions)</th>
<th>Number of new artists</th>
<th>Number of new albums</th>
</tr>
</thead>
<tbody>
<tr>
<td>1995</td>
<td>722.9</td>
<td>3822</td>
<td>7576</td>
</tr>
<tr>
<td>1996</td>
<td>778.9</td>
<td>4093</td>
<td>7855</td>
</tr>
<tr>
<td>1997</td>
<td>753.1</td>
<td>4216</td>
<td>8642</td>
</tr>
<tr>
<td>1998</td>
<td>847.0</td>
<td>4755</td>
<td>9196</td>
</tr>
<tr>
<td>1999</td>
<td>938.9</td>
<td>5472</td>
<td>10,344</td>
</tr>
<tr>
<td>2000</td>
<td>942.5</td>
<td>5883</td>
<td>11,198</td>
</tr>
<tr>
<td>2001</td>
<td>881.9</td>
<td>5920</td>
<td>11,819</td>
</tr>
<tr>
<td>2002</td>
<td>803.3</td>
<td>6005</td>
<td>12,925</td>
</tr>
<tr>
<td>2003</td>
<td>746.0</td>
<td>6851</td>
<td>14,153</td>
</tr>
<tr>
<td>2004</td>
<td>787.0</td>
<td>7931</td>
<td>15,941</td>
</tr>
</tbody>
</table>

Sales figures in the first two columns are from RIAA, and include digital music sales in 2004. Revenues are deflated to 1999 dollars using the CPI. The second column reports the number of new artists (i.e., artists releasing debut albums) each year, based on the MusicBrainz database. The third column reports the number of new albums released (by all artists) each year, also based on the MusicBrainz database.

Table 5
Concert revenue growth rate – matched sample of artists.

<table>
<thead>
<tr>
<th>Broadband penetration quartile</th>
<th>National</th>
<th>0–25%</th>
<th>25–50%</th>
<th>50–75%</th>
<th>75–100%</th>
</tr>
</thead>
<tbody>
<tr>
<td>Cohort 1 (rank 1–50)</td>
<td>0.120</td>
<td>0.104</td>
<td>0.134</td>
<td>0.118</td>
<td>0.107</td>
</tr>
<tr>
<td></td>
<td>(0.016)</td>
<td>(0.024)</td>
<td>(0.016)</td>
<td>(0.012)</td>
<td>(0.018)</td>
</tr>
<tr>
<td>Cohort 2 (rank 51–100)</td>
<td>0.073</td>
<td>0.126</td>
<td>0.066</td>
<td>0.141</td>
<td>0.074</td>
</tr>
<tr>
<td></td>
<td>(0.013)</td>
<td>(0.027)</td>
<td>(0.015)</td>
<td>(0.015)</td>
<td>(0.014)</td>
</tr>
<tr>
<td>Cohort 3 (rank 101–200)</td>
<td>0.076</td>
<td>0.159</td>
<td>0.086</td>
<td>0.142</td>
<td>0.085</td>
</tr>
<tr>
<td></td>
<td>(0.005)</td>
<td>(0.026)</td>
<td>(0.010)</td>
<td>(0.015)</td>
<td>(0.010)</td>
</tr>
<tr>
<td>Cohort 4 (rank 201–300)</td>
<td>0.093</td>
<td>0.225</td>
<td>0.168</td>
<td>0.159</td>
<td>0.123</td>
</tr>
<tr>
<td></td>
<td>(0.006)</td>
<td>(0.040)</td>
<td>(0.012)</td>
<td>(0.018)</td>
<td>(0.010)</td>
</tr>
<tr>
<td>Cohort 5 (rank 301–409)</td>
<td>0.153</td>
<td>0.139</td>
<td>0.254</td>
<td>0.201</td>
<td>0.208</td>
</tr>
<tr>
<td></td>
<td>(0.006)</td>
<td>(0.052)</td>
<td>(0.016)</td>
<td>(0.025)</td>
<td>(0.013)</td>
</tr>
<tr>
<td>Napster + Cohort 1</td>
<td>−0.054</td>
<td>0.137</td>
<td>−0.092</td>
<td>−0.100</td>
<td>−0.020</td>
</tr>
<tr>
<td></td>
<td>(0.025)</td>
<td>(0.035)</td>
<td>(0.024)</td>
<td>(0.021)</td>
<td>(0.028)</td>
</tr>
<tr>
<td>Napster + Cohort 2</td>
<td>0.066</td>
<td>0.128</td>
<td>0.068</td>
<td>−0.031</td>
<td>0.096</td>
</tr>
<tr>
<td></td>
<td>(0.017)</td>
<td>(0.041)</td>
<td>(0.017)</td>
<td>(0.024)</td>
<td>(0.021)</td>
</tr>
<tr>
<td>Napster + Cohort 3</td>
<td>0.118</td>
<td>0.219</td>
<td>0.108</td>
<td>0.097</td>
<td>0.145</td>
</tr>
<tr>
<td></td>
<td>(0.008)</td>
<td>(0.038)</td>
<td>(0.011)</td>
<td>(0.022)</td>
<td>(0.017)</td>
</tr>
<tr>
<td>Napster + Cohort 4</td>
<td>0.154</td>
<td>0.271</td>
<td>0.092</td>
<td>0.185</td>
<td>0.186</td>
</tr>
<tr>
<td></td>
<td>(0.010)</td>
<td>(0.055)</td>
<td>(0.017)</td>
<td>(0.027)</td>
<td>(0.022)</td>
</tr>
<tr>
<td>Napster + Cohort 5</td>
<td>0.180</td>
<td>0.447</td>
<td>0.069</td>
<td>0.204</td>
<td>0.181</td>
</tr>
<tr>
<td></td>
<td>(0.012)</td>
<td>(0.065)</td>
<td>(0.022)</td>
<td>(0.035)</td>
<td>(0.030)</td>
</tr>
<tr>
<td>Observations</td>
<td>3272</td>
<td>1475</td>
<td>2725</td>
<td>2295</td>
<td>2829</td>
</tr>
<tr>
<td>R-squared</td>
<td>0.524</td>
<td>0.338</td>
<td>0.423</td>
<td>0.364</td>
<td>0.366</td>
</tr>
</tbody>
</table>

Robust standard errors reported in parentheses. Analyses rely on the matched artist sample for concerts and albums. The concert revenue growth rate is calculated as the log of current year concert revenues minus the log of previous year concert revenues for the same revenue rank artist in both years based on revenues in the region of analysis. Cohorts are based on national concert revenue ranks for all analyses. Note that these regressions do not include a constant term (to make it easier to interpret and compare the coefficients across cohorts), so the $R^2$ does not have the usual interpretation.

19 Our analysis compares average growth rates in the pre- and post-Napster periods. We average over the pre- and post-years because yearly changes in growth rates are rather noisy, and show no significant trends over time.
concert revenues.\textsuperscript{20} We create identifier variables for each cohort group (e.g., for the matched concert revenue data, cohort 1 reflects nationally ranked artists 1–50, cohort 2 is 51–100, and so on). We use only year-rank observations of 409 or better because that is the lowest rank for which we maintain a balanced sample from 1995 to 2002.\textsuperscript{21} Finally, we calculate a dummy variable to identify years following the entry of Napster (i.e., 2000–2002) and interact this variable with our cohort groups.

Analyses conducted for groups of DMAs we refer to as “DMA-level” analyses. For these analyses, the growth rates are calculated based on the artists’ DMA-specific ranks (rather than the national ranks). For example, when we analyze album sales for the New York DMA, we rank artists based on their aggregate album sales in just the New York DMA and calculate album sales growth rates based on the change in the log of current album sales for a given New York DMA-ranked artist.\textsuperscript{22} While growth rates are calculated at the DMA level, artists continue to be assigned to cohorts based on their national aggregate concert revenues or album sales. Calculating growth rates based on ranks that correspond to the DMA, but assigning cohort groups based on national outcomes addresses two issues. First, using the rank for the DMA to calculate growth rates ensures that

\begin{table}[h]
\centering
\begin{tabular}{lcccc}
\hline
& \multicolumn{4}{c}{Broadband penetration quartile} \\
& \multicolumn{1}{c}{National} & \multicolumn{1}{c}{0–25\%} & \multicolumn{1}{c}{25–50\%} & \multicolumn{1}{c}{50–75\%} & \multicolumn{1}{c}{75–100\%} \\
\hline
Cohort 1 (rank 1–100) & 0.098 & 0.118 & 0.092 & 0.074 & 0.093 \\
& (0.004) & (0.007) & (0.007) & (0.007) & (0.008) \\
Cohort 2 (rank 101–300) & 0.087 & 0.125 & 0.086 & 0.069 & 0.087 \\
& (0.001) & (0.006) & (0.004) & (0.005) & (0.005) \\
Cohort 3 (rank 301–500) & 0.086 & 0.115 & 0.108 & 0.082 & 0.091 \\
& (0.000) & (0.008) & (0.005) & (0.004) & (0.006) \\
Cohort 4 (rank 501–700) & 0.091 & 0.112 & 0.152 & 0.097 & 0.091 \\
& (0.000) & (0.009) & (0.005) & (0.007) & (0.008) \\
Cohort 5 (rank 701–900) & 0.108 & 0.109 & 0.177 & 0.123 & 0.074 \\
& (0.000) & (0.010) & (0.007) & (0.008) & (0.010) \\
Cohort 6 (rank 901–1100) & 0.110 & 0.135 & 0.192 & 0.122 & 0.067 \\
& (0.000) & (0.014) & (0.009) & (0.011) & (0.012) \\
Cohort 7 (rank 1101–1300) & 0.127 & 0.128 & 0.230 & 0.135 & 0.065 \\
& (0.000) & (0.016) & (0.012) & (0.012) & (0.015) \\
Cohort 8 (rank 1301–1500) & 0.141 & 0.187 & 0.229 & 0.132 & 0.059 \\
& (0.001) & (0.025) & (0.013) & (0.016) & (0.017) \\
Cohort 9 (rank 1501–1716) & 0.224 & 0.080 & 0.335 & 0.209 & 0.049 \\
& (0.005) & (0.023) & (0.025) & (0.020) & (0.024) \\
Napster + Cohort 1 & 0.028 & 0.029 & 0.030 & 0.008 & 0.004 \\
& (0.005) & (0.012) & (0.010) & (0.013) & (0.011) \\
Napster + Cohort 2 & 0.032 & 0.033 & 0.053 & 0.057 & 0.052 \\
& (0.001) & (0.010) & (0.006) & (0.007) & (0.011) \\
Napster + Cohort 3 & 0.072 & 0.135 & 0.107 & 0.165 & 0.112 \\
& (0.002) & (0.012) & (0.007) & (0.012) & (0.015) \\
Napster + Cohort 4 & 0.104 & 0.202 & 0.127 & 0.226 & 0.163 \\
& (0.001) & (0.015) & (0.010) & (0.015) & (0.022) \\
Napster + Cohort 5 & 0.130 & 0.244 & 0.154 & 0.270 & 0.223 \\
& (0.001) & (0.018) & (0.013) & (0.022) & (0.029) \\
Napster + Cohort 6 & 0.170 & 0.317 & 0.220 & 0.358 & 0.356 \\
& (0.001) & (0.025) & (0.015) & (0.026) & (0.033) \\
Napster + Cohort 7 & 0.186 & 0.313 & 0.279 & 0.344 & 0.420 \\
& (0.001) & (0.024) & (0.022) & (0.029) & (0.041) \\
Napster + Cohort 8 & 0.211 & 0.316 & 0.274 & 0.423 & 0.479 \\
& (0.000) & (0.035) & (0.024) & (0.039) & (0.051) \\
Napster + Cohort 9 & 0.176 & 0.512 & 0.282 & 0.391 & 0.497 \\
& (0.004) & (0.037) & (0.034) & (0.045) & (0.064) \\
Observations & 17,160 & 8089 & 9762 & 9473 & 11,046 \\
R-squared & 0.578 & 0.409 & 0.570 & 0.417 & 0.208 \\
\hline
\end{tabular}
\caption{Concert revenue growth rate – pollstar dataset.}
\end{table}

Robust standard errors are reported in parentheses. Analyses rely on the all artists in the Pollstar concert data. The concert revenue growth rate is calculated as the log of current year concert revenues minus the log of previous year concert revenues for the same revenue rank artist in both years based on revenues in the region of analysis. Cohorts are based on national concert revenue ranks for all analyses. Note that these regressions do not include a constant term (to make it easier to interpret and compare the coefficients across cohorts), so the $R^2$ does not have the usual interpretation.

\textsuperscript{20} If we define cohorts based on album sales, we get results that are very similar to those reported in the tables below.

\textsuperscript{21} There are 409 artists with positive concert revenues in 1993 in our matched sample, and this count of artists increases for every year following 1993. Similarly, there are 703 artists with album sales in 1993 in our matched sample, and this count of artists also increases for every year following 1993. To ensure that annual concert revenues (album sales) are not overly impacted by the growth in the number of artists performing (recording) during the course of the year we drop 1993 from the analyses. Using the remaining data we can then calculate growth rates for concert revenues (album sale) starting in 1995.

\textsuperscript{22} When several DMAs are represented in a given area, such as DC and Baltimore, we aggregate concert revenues and album sales for each artist across all of the DMAs in that area, and then calculate ranks and growth rates based on the aggregated data.
growth rates are based on artists who have a similar level of popularity in a given DMA (i.e., the 10th-ranked artist in the DMA this year compared to the 10th-ranked artist in the DMA last year) and results in smoother and more reasonable growth rates. Second, relying on the national ranks to assign artists to cohorts ensures that the cohorts and their corresponding coefficients are comparable across analyses.\textsuperscript{23}

4.1. National and DMA-level analyses

Table 5 reports results of analyses on concert revenue growth rates for the matched sample of artists. To make it easier to interpret and compare the regression coefficients across cohorts, so the $R^2$ does not have the usual interpretation.

For example, it may not be appropriate to compare the top ranked concert in the Albuquerque DMA to the top ranked concert in the New York DMA.

\textsuperscript{23} For example, it may not be appropriate to compare the top ranked concert in the Albuquerque DMA to the top ranked concert in the New York DMA.
incidence of file-sharing, one would expect to find that the impact of file sharing on concert revenues and album sales was more pronounced for DMAs with greater broadband penetration. Instead, the results appear to be quite similar across all of the quartiles for broadband penetration and similar to the national results. One possible explanation is that CD buyers (post-Napster) are a different consumer segment than concert-goers and file-sharers. A related explanation is that our measure of broadband penetration covers each DMA as a whole, and ideally one would like a measure of broadband penetration for the demographic in each DMA that is most likely to attend concerts and purchase albums. Unfortunately, such a measure is not available. A further complicating factor is that the subpopulation that most frequently attends concerts and purchases albums may be early to adopt broadband access regardless of their DMA (e.g., college students). Finally, variation in broadband penetration across DMAs for the relevant sub-population that attends concerts and purchases albums may be quite small. If this is in fact the case, then we would expect to find little difference in the impact of file-sharing across DMAs on concert revenue or album sales.

Since the set of artists in our matched sample is a subset of those who performed concerts during the sample period, we repeated the analyses described above for the full Pollstar sample. The results, which are reported in Table 6, are very similar. We also estimated the regression described in Table 5 at the DMA level separately for the three largest DMAs (New York, Los Angeles, and Chicago). While the levels of the coefficients differ somewhat, the comparisons across cohorts are consistent with the national-level results in column 1 of the table. In each case, the concert revenue growth rate for cohort 1 (the largest artists) declines following the entry of Napster, while concert revenue growth rates for smaller artists increase significantly following entry of Napster.

Table 7 reports results for the album sales analyses. These analyses are constructed in a comparable manner as those for concert revenues in Table 5, but artists are instead ranked based on album sales and growth rates are calculated using album sales. All cohorts experience a decline in album sales following entry of Napster, which is consistent with the findings in prior papers. However, the decrease in album sales is sharpest for the largest artists (cohort 1). In the national level analyses, the coefficient on the Napster interaction is $0.171$ and indicates a significantly lower album sale growth rate for cohort 1 following entry of Napster compared to before Napster (when the growth rate was $0.077$). All of the smaller artist cohorts also experienced lower growth rates following Napster, but the changes were more muted for most of the smaller artists. The differences between the interaction coefficients for cohorts 2–8 vs. cohort 1 are all statistically significant. As with the results for concert revenue growth rates, the results using album sales growth rates suggest that file sharing may have increased awareness of smaller artists and boosted demand for small artists’ concerts and albums.
This is consistent with the hypothesis that the positive demand implications of file sharing should be much stronger for small, more obscure artists than for larger, better-known artists. As with concert revenues, the results are similar across different quantiles of DMAs based on their broadband penetration. Results for the three largest DMAs (New York, Los Angeles, Chicago), and all other DMAs are also consistent with the national results.

4.2. Robustness checks

We explore robustness of the main results for concert revenues in Table 8. The first column of the table repeats the national-level results from Table 5 for purposes of comparison. In columns 2 and 3, we control for underlying time trends in concert revenue growth rates. As mentioned above, we found no evidence of significant time trends prior to 1999, so it is not surprising that our main results are robust to the inclusion of a time trend. In the last column of the table, we re-estimate our main specification using log concert revenues as the dependent variable (instead of the growth rate). The results again confirm that revenues increased much more sharply for small artists than for large artists.

Table 9 reports analogous results for album sales. When growth rates are calculated relative to a linear or quadratic time trend (columns 2 and 3), the magnitudes of the Napster coefficients shift, but the comparisons across cohorts are roughly the same as in our baseline specification: relative to trend, sales fell most rapidly (or grew most slowly) for large artists. If we estimate the same specification using log sales as the dependent variable, the results confirm the basic finding that post-Napster sales were weakest for large artists.

In addition to the results reported in Tables 8 and 9, we also experimented with specifications that are flexible on the timing of the shock. In particular, if we include year dummies and interactions with year dummies (instead of simply a post-Napster dummy), the timing implied by the estimated coefficients roughly matches the pattern shown in Fig. 1. Concert revenues and album sales change significantly in 2000, with the largest changes coming in 2001 and 2002.
5. Discussion and conclusion

We find evidence consistent with illegitimate redistribution of digital goods increasing revenue from non-digital complementary products. As with the earlier literature, we find that sales of recorded music declined precipitously with the entry of Napster and large-scale file-sharing. While file-sharing may have substantially displaced album sales, it also facilitated a broader distribution of music, which appears to have expanded awareness of smaller artists and increased demand for their live concert performances. Concert revenues for large artists, however, appear to have been largely unaffected by file-sharing. Music for large artists was likely widely available prior to file-sharing, and as a result it is not surprising that demand for those artists’ concerts would have been largely unaffected by file-sharing. Similarly, the decline in album sales is much more pronounced for large artists than for small artists. Large artists’ music may have been more readily available on file-sharing sites, in which case file-sharing would naturally have a bigger impact on large artists. But for small artists, file-sharing may have increased awareness of their music and encouraged some additional album sales from a larger fan base even as it displaced album sales to others. While the market for live music appears to expand after Napster, and the market for recorded music contracts, the results imply that large artists lose market share in both markets.

While our findings are consistent with file-sharing affecting concert revenues and album sales, we cannot entirely rule out the influence of other contemporaneous changes in the music industry. During the time period of our analysis, the concert promotion industry became increasingly concentrated, with Clear Channel Entertainment (now Live Nation) gaining an increasingly large share of concert promotion activities. At the same time, Clear Channel was also expanding the breadth of radio stations under its control. Finally, digitization not only affected the reproduction and redistribution of recorded music, but also changed production technologies more generally. While all of these factors could have influenced growth rates for concert revenues and album sales, we would expect their influence to have been more gradual and more consistent across large and small artists than what we observe in the data. Given the sharp changes that we observe over a very short period of time around the entry of Napster, along with the differences in the effects for large vs. small artists, the findings in this paper are more easily explained by file-sharing than by other contemporaneous factors.

More generally, increased digitization of information and entertainment content over the past decade, along with dramatic technological changes in the reproduction and redistribution of such goods, has raised substantial concerns with respect to the future viability of many information and entertainment goods markets. Recorded music, newspapers, and magazines have all undergone radical changes through on-line distribution and pricing that have either facilitated illegitimate redistribution of their content or generally increased the availability of legitimate but free content. Markets for television, books and movies have just recently started to embrace digital distribution channels, but have also faced challenges from free and/or illegitimate distribution of their content (e.g., YouTube). Finally, software has long fought against illegal copying and is also taking some steps toward free legitimate distribution through "cloud" computing rather than personal licensing. These changes are undoubtedly having profound impacts on the market structure of these industries, making it more difficult to generate revenue from traditional sources but also greatly expanding overall distribution and availability of content. To the extent that content in these industries becomes available to a larger potential customer base, some of the decline in revenue from traditional sources may be offset by increased demand for complementary products.

Acknowledgements

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References