# Measuring the Impact of Radio Play on Record Sales 

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#### Abstract

: Received wisdom states that radio play enhances sales of sound recordings. Nevertheless, the relationship between radio play and the sales of sound recordings has received almost no empirical analysis. This paper undertakes an econometric investigation of the impact of radio play on sales of sound recordings using a sample of American cities. The results indicate that radio does not have a positive impact on record sales. In fact, although measured somewhat imprecisely, radio play appears to have an economically important negative economic impact, implying that radio is more of a substitute for records than a complement. This research exposes an important fallacy of composition in applying to an entire market the obvious positive relationship between radio broadcasts and record purchases that holds for individual songs. Radio airplay is properly understood to be capable of rearranging the record purchases made by consumers while at the same time shrinking the overall market for sound recordings.


The economic impact of radio play on the sound recording market has received scant attention from the academic community. ${ }^{1}$ Y et radio broadcasts and sound recordings are intimately intertwined. The majority of radio stations, for example, are primarily described by the genre of sound recordings that they broadcast, whether it is 'classic rock', 'hot adult contemporary', or 'cool jazz'.

It is clear that radio broadcasters can have a profound impact on the success of individual sound recordings and are often paid by record companies to help achieve that goal. ${ }^{2}$ It is possible that this fact has led potential researchers to conclude that radio play has a beneficial impact on overall record sales, seemingly obviating the need to perform an analysis. Nevertheless, radio's positive impact on individual recordings may not hold for the industry as a whole. As discussed below there is a potentially important fallacy of composition in this market.

Nor can we discover the impact through direct observation. O ne might expect that we could discover the overall impact of radio play merely by examining the market transactions that occur between these parties. We know, for example, that television broadcasters place higher values on movie broadcasts than do movie owners (for whom it is probably negative in terms of DVD sales). We know this because television stations pay positive prices for the rights to broadcast movies. ${ }^{3}$ If there were a similar market for rights to broadcast music over radio we would know the impact of radio play by direct observation - either broadcasters would pay the sound recording owners or the sound recording owners would pay the broadcasters. But there is no such market-owners of sound recordings in the

[^0]United States (and other countries) do not have the legal ability to restrict the broadcasts of their recordings. ${ }^{4}$

Recent technological trends, especially a new generation of digital radio broadcast receiversterrestrial, satellite, and Internet based that are capable of making and storing copies of sound recordings - appear likely to increase the visibility and importance of this issue. These receivers alter the typical "streaming" nature of radio, which has historically broadcast songs whose only trace remained in the memory of the listener. The new receivers allow users to automatically record digital songs, providing unlimited playback at the discretion of the user. There have already been several recent skirmishes between these industries and we can expect more friction as these technologies mature. ${ }^{5}$ This would seem, therefore, to be a propitious time to examine the nature of this interaction.

## I. A Brief History of Radio and Sound Recording

Radio and sound recordings have largely grown up together with both industries reaching commercial viability early in the $20^{\text {th }}$ century, although sound recordings came first. Thomas Edison is credited with creating the first sound recording in 1877 with a tinfoil recording process. Tinfoil was soon replaced with wax cylinders, leading to a long-forgotten standards battle between cylinders and disks (the disk system, known as the gramophone was developed by Emile Berliner). Just as VHS came later but nevertheless won its battle with Beta, disks came later but eventually won the day.

The first commercial American radio stations went on the air in late 1920. Numerous stations were borne in the next few years and by 1923 the number of stations was over 500 , which remained the

[^1]approximate number for the next fifteen years (Hazlett 1997). ${ }^{6}$ In 1926 the penetration rate of radio was approximately $20 \%{ }^{7}$ In those days both radio and sound recordings were more the provenance of the middle and upper classes than the lower class and the overall penetration rates probably underestimate the penetration rate of radio in sound recording households.

The market for sound recordings was surprisingly mature by the time of radio's entrance. For example, a magazine devoted to the sound recording industry (Talking Machine World) was established in 1905 and by 1920 monthly issues were averaging 200 pages in size. ${ }^{8}$ Sound recording sales in 1921 were more than $\$ 1.1$ billion, measured in 2004 dollars and the population was only slightly more than one third of the current population. ${ }^{9}$ To put this value in perspective, constant dollar sales revenue per capita was actually slightly higher in 1920 than in 1950.

Liebowitz (2004) examined the historical relationship between record sales and radio play for two periods: the introduction of radio in the US in the 1920s and the introduction of commercial radio in Britain in the latter decades of the $20^{\text {th }}$ century. In the first instance record sales fell dramatically after the introduction of radio, and in the second case there was no evidence of a positive relationship between new radio play and record sales. The current paper is an attempt to more directly and more precisely measure the relationship between radio play and sound recordings.

## II. The Possible Relationships between Radio and Sound Recordings

It is often claimed that radio has a beneficial impact on sound recording sales. While it is incontrovertible that radio can direct demand to particular songs that receive heavy airplay, the impact

[^2]on individual songs is quite distinct from the impact on the entire industry, although this distinction has not been generally recognized.

The particular details of the overall impact of radio depend on two competing factors. On the one hand, radio allows users to experience new songs that they may not have previously heard. If this were the primary use of radio by listeners then radio could easily increase overall record sales. On the other hand, the time spent listening to radio is also capable of being a substitute for the time spent listening to prerecorded music. To the extent that broadcast radio is such a substitute, radio would be expected to harm overall record sales. Radio is capable of delivering both impacts and the relative strength of each would determine the overall impact.

## A. What can we learn from statistics on music listening?

The bare statistics on time spent listening to music by source are informative in and of themselves. The average American spent five times as much time listening to radio per day than listening to traditional sound recordings in 2003, according to the US Statistical abstract. ${ }^{10}$ These timeusage values seem incompatible with a hypothesis that radio is used primarily as a means to learn about new music for later purchase, since it would appear infeasible that consumers spend so much more time searching for new music then they spend in the ultimate act of music consumption. These statistics imply that radio is being used largely for its own consumption value.

Certainly, this line of thinking doesn't prove that time spent listening to radio is too long to be pure search, but it illustrates the great likelihood that much and probably most radio listening is a form of consuming music, and if so, radio is likely to be a substitute for the listening to and the purchasing of

[^3]sound recordings. Understanding the nature of that substitution depends on understanding the nature of music consumption.

## B. Music Consumption

Listening to music is a favorite activity for many individuals. The particular forms of consumption are varied, however, and include attending live performances, listening to CDs (or other sound recording mediums), or listening to radio and television broadcasts. Our focus is on the two major sources- broadcast radio and sound recordings. These two music sources satisfy the music listening craving in different ways and each has certain advantages relative to the other.

Sound recordings provide the highest audio quality and also allow particular songs and performances to be ideally matched to the individual's tastes. Broadcast radio, besides suffering from lower audio quality and less perfectly matched music, also suffers from numerous minutes of advertising. Nevertheless, radio has some advantages over sound recordings— disk jockey patter (which many consumers apparently enjoy); broad playlists which allow the consumer to sit back and let someone else decide what to play (which is presumably more useful than a pure randomizer switch since otherwise radio would just use such a switch; and a much lower price since radio is free whereas the legal consumption of sound recordings requires that they be purchased.

These different characteristics provide different strengths in catering to the music listening desires of consumers. We can think of two extremes in a continuum of music listening experiences. On the one hand, an individual might wish to listen to a specific recorded performance or set of performances, which we can refer to as "specific" music consumption. Alternatively, an individual might wish to listen to a random selection of performances from a large library of performances (most likely from a particular genre) which we can refer to as generic or nonspecific music consumption. The
two types of listening, which are themselves somewhat substitutable, imply different behavior toward radio and sound recordings.

If specific music consumption is desired the individual will need to access the specific sound recordings of interest, either from his personal collection, those of acquaintances, or perhaps libraries. Once these sound recordings are in the individual's possession, he can easily and quickly listen to the songs in which he is most interested. Radio, by way of comparison, is not an efficient technology for accessing specific songs. Since a song is considered to be in heavy rotation if it is played twice a day, an individual would need to spend an inordinate amount of time listening to radio before even one desired song was played, to say nothing of a larger collection of songs (note that this is somewhat less true for satellite radio which sometimes has a station devoted to songs from but a single artist, e.g., the Elvis Presley or Bruce Springsteen stations on Sirius Satellite Radio).

Non-specific music consumption is another matter entirely. Radio is particularly good at catering to this desire, with its playlists and large libraries. Individuals can use their personal libraries to also provide a form of non-specific listening, perhaps by telling their CD or MP3 player to randomize the play of songs, or else choosing the music to listen to in a somewhat haphazard manner. Because sound recordings are not free, the music libraries of individuals are usually quite limited in comparison to that of radio stations. The disadvantages of radio are its lower audio quality and the fact that its collection of music is not as closely tailored to the tastes of individual listeners as their own libraries are likely to be. Nevertheless, the relative usage statistics reported above indicate that the disadvantages of radio are overwhelmed by its advantages for a great majority of individuals.

Note that radio and sound recording are substitutes for non-specific music consumption whereas specific music consumption should be dominated by the use of sound recordings. More importantly, radio broadcasts are clearly a substitute for sound recordings in the case of non-specific music
consumption but may well be a complement for sound recordings in the specific music consumption category. This latter result is due to the fact that radio can provide information and therefore influence which specific sound recordings are purchased.

Because radio and sound recordings compete for non-specific music uses, radio usage will have negative impacts on the sales of sound recordings for non-specific music uses, which appears to be by far the larger of the two uses. In the much smaller category of specific music use, radio will clearly influence the selection of sound recordings and may even increase the number of sound recordings sold. By focusing on the latter interaction of these music sources to the exclusion of the former interaction, previous discussion have ignored the potentially negative impact of radio on sound recording sales. We turn now to an empirical investigation of that relationship.

## III. Data

In order to perform our analysis we need to merge three data sets together: Arbitron data on radio, Nielsen SoundScan data on record sales, and US Census data for market demographics.

The Arbitron radio data are based upon diaries filled out by respondents, similar to Nielsen television diaries. The data are produced several times a year and currently are found in digital form. We were provided access to their data for 1998 and 2003. Arbitron classifies stations by type and also aggregates groups of stations into approximately 275 (269 and 278 in 1998 and 2003 respectively) Metropolitan Survey Areas (known as Metro Areas) based on the areas in which they broadcast. Some rural residents are left out of the surveys. Arbitron data include information on the average time spent
listening to radio in its Metro Areas as well as data on the share and genre of each radio station in an area, allowing a calculation to be performed separating the audiences for music radio and talk radio. ${ }^{11}$

Nielsen SoundScan sells data on record sales (full length albums), by geographic area, genre, and by year. Sales data come mainly from bar code scanners at retail outlets. Online sales are included in these numbers, with customer locations mapped to shipping addresses for physical units or credit card locations for digital downloads. As a factual matter, digital downloads played virtually no role in the analysis since they were a trivial component of the market even as late as 2003. We purchased data for the largest 100 areas, which Nielsen refers to as Designated Market Areas (DMAs) of which there are 210 in the US. Nielsen's DMAs include everyone in the United States. The 100 largest DMAs include approximately $83 \%$ of the total population.

The US Census, as part of it Current Population Survey (CPS) undertaken for the Bureau of Labor Statistics, conducts irregular surveys on Internet and Computer use. We use these Census surveys since we wish to control for the important impact of file-sharing on record sales. There was a survey in December of 1998 and another in October of 2003 and these are the two used in the analysis. ${ }^{12}$ The surveys provide information on demographic variables such as average household income, age distribution by area, minority share of population, breakdown by gender, internet use, type of internet connection, as well as a host of other variables not used in the analysis. The geographic areas used in the Census are known as Metropolitan Statistical Areas (MSAs) and there are 241 of these areas in our data. As is the case with Arbitron Metro Areas, these MSAs do not include rural residents. ${ }^{13}$ Census data are based on responses from individuals to survey questions. The size of the census survey

[^4]sample (approximately 130,000 nationally) in small MSAs is sometimes insufficient to provide accurate estimates for various demographic data. We try to take account of this problem in the analysis. Arbitron Metro Areas normally correspond to Census MSAs although they are not identical to them. ${ }^{14}$

Combining these data sets is not a trivial task. Since Nielsen DMAs are the largest areas and represent larger populations than Census MSAs or Arbitron Metro Areas (even when they all have the same name) we aggregated the MSAs and Metro Areas to match the Nielsen DMAs. This often required adding several MSAs (or Metro Areas) together to approximate the D MA. Arbitron provides a guide to link its Metro areas to the Nielsen DMAs, although the resulting matches are sometimes far from perfect. Matching the Census MSAs to the Nielsen DMAs was based upon examining Nielsen DMA maps (which show the counties belonging to a DMA) and determining which DMA an MSA belonged to based on the county containing the MSA.

The 'matched' Metro Areas and Census MSAs sometimes contained only a small portion of the D MA population, particularly for the D MAs with smaller populations and more rural characteristics. This is because rural households in DMAs are often excluded from Metro Areas and MSAs. For that reason we constructed a variable, "Coverage", which measures the portion of the DMA population replicated by the aggregated MSAs or Metro Areas. ${ }^{15}$ When Coverage falls to a low level it is possible that the Census or Arbitron variables, based as they are on MSAs which make up only a small percentage of the DMA population, will not properly reflect the actual population characteristics in the D MA. In the analysis that follows the sample will sometimes be restricted to observations where the

[^5]Coverage is greater than $60 \%$ or $75 \%$, in order to eliminate the influence of potentially misleading measurements.

Although the data from Nielsen SoundScan cover 100 DMAs, one DMA could not be matched with any census MSAs and was dropped from the analysis. Further, missing data for radio listenership removed another three or four DMAs, depending on year and whether radio was measured as total radio audience or music radio audience.

| Table 1: 2003 Values |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Variable | Obs | Mean | Std. Dev. | Min | Max | weighted | Rural |
| Household Income (000) | 99 | 47.966 | 8.986 | 20.380 | 75.895 | 50.540 | 38.255 |
| College Degree | 99 | 0.204 | 0.051 | 0.087 | 0.345 | 0.216 | 0.139 |
| DMA Population (00000) | 99 | 23.505 | 27.275 | 6.308 | 194.212 | 54.835 |  |
| Males | 99 | 0.480 | 0.023 | 0.400 | 0.520 | 0.482 | 0.484 |
| Coverage | 99 | 0.683 | 0.206 | 0.203 | 0.977 | 0.828 |  |
| Minority | 95 | 0.220 | 0.138 | 0.024 | 0.665 | 0.269 | 0.293 |
| Old (55+) | 99 | 0.227 | 0.054 | 0.130 | 0.410 | 0.215 | 0.250 |
| Radio Usage | 96 | 2.711 | 0.161 | 2.371 | 3.233 | 2.769 |  |
| Music Radio Usage | 96 | 2.298 | 0.190 | 1.861 | 2.976 | 2.293 |  |
| Talk Radio Usage | 95 | 0.417 | 0.138 | 0.190 | 0.750 | 0.476 |  |
| Record Sales per capita | 99 | 2.321 | 0.440 | 1.499 | 3.879 | 2.445 | 1.837 |
| Share Internet | 99 | 0.613 | 0.071 | 0.440 | 0.740 | 0.621 | 0.545 |
| Young (12-29) | 99 | 0.303 | 0.044 | 0.200 | 0.410 | 0.306 | 0.288 |
| Number Radio Stations | 95 | 22.017 | 4.991 | 12.287 | 38.109 | 25.304 |  |
| Weight | 99 | 651.593 | 545.538 | 17.108 | 2664.062 |  |  |
| 98-2003 |  |  |  |  |  |  |  |
| Variable | obs | Mean | Std. Dev. | Min | Max |  |  |
| radiochg | 95 | -0.294 | 0.104 | -0.600 | -0.050 |  |  |
| radiomuschg | 95 | -0.323 | 0.123 | -0.623 | -0.036 |  |  |
| salpercapchg | 99 | -0.577 | 0.695 | -3.484 | 1.049 |  |  |
| incomechg | 99 | 8.523 | 7.087 | -6.660 | 26.901 |  |  |
| yngchg | 99 | 0.001 | 0.045 | -0.110 | 0.140 |  |  |
| dmapopchg | 99 | 1.643 | 2.361 | -0.559 | 13.845 |  |  |
| shrmlchg | 99 | 0.001 | 0.035 | -0.137 | 0.143 |  |  |
| netusechg | 99 | 0.310 | 0.058 | 0.120 | 0.466 |  |  |
| oldchg | 99 | 0.011 | 0.047 | -0.120 | 0.191 |  |  |
| collegech | 99 | 0.018 | 0.040 | -0.114 | 0.208 |  |  |
| radtalkchg | 95 | 0.029 | 0.092 | -0.227 | 0.351 |  |  |
| inchgthous | 99 | 8523.070 | 7087.131 | -6660.040 | 26901.410 |  |  |
| radiom03 | 95 | 2.296 | 0.178 | 1.861 | 2.976 |  |  |
| stacntchg | 96 | 2.172 | 7.311 | -11.404 | 65.000 |  |  |
| minoritychg | 93 | 0.019 | 0.054 | -0.115 | 0.186 |  |  |

Table 1 presents summary statistics for 2003 and for the change from 1998 to 2003, allowing the reader to infer the 1998 statistics if desired. A person in the average DMA spent 2.3 hours per day
listening to music radio and 2.71 hours a day listening to all radio. Sales of full length sound recording albums averaged 2.32 per person per year across DMAs, somewhat less than the average weighted by population. The combined coverage ratio in the average DMA was $68.3 \%$ and the DMA with the lowest values was about $20 \%$, which would be a cause for concern if these observations were accorded much weight in the analyses. The national (weighted) coverage ratio was a more reassuring $82.8 \%$, however. Note that the variation in values for the small cities is likely to be overstated due to the small samples in those cities.

As mentioned, the population of the top 100 DMAs represents about $83 \%$ of the national population. The MSA (Metro Area) population matched to the DMAs covers about 87\% (79\%) of the DMA population, so that in total our sample covers about $72 \%(66 \%)$ of the US population. How does the population left out of MSAs compare to the included population? Being more rural, the left out population would be expected to be poorer, have lower Internet usage, and lower education. This expectation is confirmed in the rightmost column of Table 1 where we see that left out individuals have lower Internet use, fewer college degrees, lower incomes, and lower record sales than the included population.

## IV. Estimation

Our goal is to determine the impact of radio play on record sales. Our null hypothesis will be that radio increases record sales since that conclusion seems to have been accepted by almost everyone. All of our variables are measured as the average value in a city. The dependent variable will be record sales per capita. The key independent variable will be the average time spent listening to music radio. Demographic variables that are likely to influence record sales include income, Internet use, possession of college degree, relative size of age groups (over 55 and 12 through 29), and minority population (black and Hispanic).

Note that Internet access can be a proxy for file-sharing and file-sharing decreases record sales (see Liebowitz 2006, Rob and Waldfogel 2006, and Zentner 2006). Nevertheless, Intemet use might also have a positive association with record sales since those individuals who have purchased computers and Internet access are likely to have greater interest in stereo systems and sound recordings. Any impact of Internet use file-sharing will be a mixture of these two impacts in 2003 but not in 1998, which predates file sharing.

We have data for 1998 and 2003. Having data for more than one year allows panel methods to be used and this will be our preferred methodology. The appendix present results from the single-year cross section regressions.

## A. Panel Data

Table 3 presents results from using first differences. By taking first differences we control for underlying differences in the populations and circumstances of cities that do not change over this period and for which we do not have controls, in a manner identical to a fixed effects model.

The table includes regression results over the full 1998-2003 interval where all the variables are in first differences, except for the measurement of Internet usage which will be explained shortly. The dependent variable is the change in albums sold per capita. The various specifications in Table 2 differ from one another as we stratify the observations by coverage ratio and population in order to remove less precisely measured observations from the analysis.

The first column includes the full sample although the results are likely to be highly influenced by poor measurements. The second column weights each observation by a combination of population and coverage, so that larger cities are more heavily weighted and cities with greater coverage are more heavily weighted, with the weighting constructed to give approximately equal impact to population and
coverage. ${ }^{16}$ The purpose of this weighting was to reduce the impact of observations with likely mismeasurement due to low coverage or possible imprecision in the Census numbers because the sample size was too small to provide reliable statistics. The weighting here is quite severe, with the variation from the highest to lowest weight on the order of over one hundred to one (seen in Table 1). The next two columns eliminate observations (giving them a zero weight) when the coverage is less than either $60 \%$ or $75 \%$. These cutoffs were chosen as fairly natural indicators of good if not great coverage and more demanding cutoffs would have lowered the number of observations further than deemed prudent, although we will explore the impact of choosing different cutoffs later in the paper. Columns 5 and 6 add in a cutoff for population as well as coverage.

| Table 2: First Differences Regression on Change in Album Sales |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Change in | Full Sample | Pop \& Cov Wgt | Coverage $>.6$ | Coverage $>.75$ | $\begin{aligned} & \mathrm{Cov}>.6 ; \\ & \text { pop>. } 6 \mathrm{M} \end{aligned}$ | $\begin{gathered} \mathrm{Cov}>.75 ; \\ \text { pop>.6M } \end{gathered}$ |
| Daily Per Capita Music Radio (Hours) | -0.0745 | -0.7903 | -0.7507 | -1.1817 | -0.6049 | -0.7767 |
|  | (0.462) | (0.076) | (0.169) | (0.126) | (0.067) | (0.056) |
| Average Household Income (000s) | 0.0087 | 0.0227 | 0.0299 | 0.0368 | 0.0148 | 0.0220 |
|  | (0.362) | (0.025) | (0.047) | (0.086) | (0.118) | (0.034) |
| 2003 Internet Access | -1.5582 | -2.7630 | -3.4950 | -4.5426 | -2.7686 | -2.5656 |
|  | (0.185) | (0.012) | (0.043) | (0.062) | (0.003) | (0.014) |
| BA Degree or above | 3.1199 | 4.0142 | 6.2029 | 9.0215 | -3.2295 | 0.3713 |
|  | (0.162) | (0.172) | (0.081) | (0.080) | (0.188) | (0.863) |
| Share 12-29 | 5.3332 | 5.2812 | 9.0277 | 8.2210 | 0.6868 | 0.8054 |
|  | (0.077) | (0.094) | (0.022) | (0.108) | (0.792) | (0.676) |
| Share Males | -0.8486 | -2.4070 | -4.6742 | -4.9393 | 1.1555 | -0.4517 |
|  | (0.721) | (0.329) | (0.159) | (0.196) | (0.452) | (0.774) |
| Share 55+ | 1.3197 | 1.1857 | 4.9417 | 1.0563 | -0.5910 | -1.2845 |
|  | (0.568) | (0.581) | (0.144) | (0.784) | (0.775) | (0.413) |
| Share Minority | -1.0790 | -0.2796 | 0.4427 | -0.9315 | 0.6420 | -0.4186 |
|  | (0.475) | (0.844) | (0.806) | (0.700) | (0.675) | (0.744) |
| DMA Population (\%) | -0.3810 | -0.3324 | -0.4518 | 0.0504 | -0.8576 | -0.4557 |
|  | (0.684) | (0.668) | (0.663) | (0.973) | (0.154) | (0.428) |
| Constant | 0.2827 | 0.6820 | 0.9922 | 1.4393 | 1.0931 | 0.7715 |
|  | (0.719) | (0.308) | (0.342) | (0.326) | (0.050) | (0.145) |
| Observations | 90 | 90 | 61 | 41 | 53 | 36 |
| R-squared | 0.14 | 0.20 | 0.33 | 0.37 | 0.25 | 0.36 |

Robust p values in parentheses; p value for music radio is for one tail test; bold is sig at $10 \%$ level; bold underlined at 5\%, bold double underline $1 \%$

[^6]Our primary interest is in the coefficients on music radio use. The coefficients are always negative and generally allow us to reject the hypothesis that radio play increases the sales of CD s because the coefficients are near or over the border of statistical significance. The average coefficient (excluding the full sample) is -.82 . but we will round this value down to -.75 in part because when the impact of influential observations is reduced the average coefficient falls to -. 68 and in part because it is a common fraction. ${ }^{17}$ This value tells us that a one-hour increase in usage of music radio, which is large relative to the average value, would lead to a decline of .75 sound recordings. Although these coefficients are not measured with enough precision to us to reject the possibility of zero impact with typical levels of statistical significance, their size implies an important economic impact of radio play on record sales since the per capita purchases of sound recordings is about 2.7 per year over the five year interval. If this coefficient could be applied to the entire range of radio usage, and we will have more to say about this below, the decline in record sales would be very large relative to actual sales.

Income is always positive and usually significant. An increase in income of $\$ 10,000$ would lift sound recording sales by approximately .25 units. DMA population has no clear impact on sales.

The Internet variable requires some additional explanation. In the period from 1998 until 2003 file-sharing arose from nothing to become a very popular activity. Liebowitz (2006) demonstrated that a correct specification for a regression hoping to measure the impact of file-sharing, if file-sharing was zero in the beginning period, would be to use the level of Internet use in the later period in an otherwise first differenced regressions. As was the case in that paper, the Internet variable in Table 2 indicates a very strong negative impact of file sharing on record sales. The full impact of file-sharing is less than this coefficient, however, because Internet usage itself is something of a substitute for listening to

[^7]sound recordings as described in Liebowitz (2006), which attempts to control for this factor and concludes that file-sharing still has a large negative impact on record sales.

The share of the population with college degrees appears to have a positive impact on record sales until small cities are removed. When influential observations were made less influential, this variable became clearly insignificant. The minority and age group variables do not have much consistency or explanatory power. The coefficient on share of individuals aged 12-29 appears to have a positive impact on record sales, but as was the case with the college variable, the result goes away when small cities are removed or when robustness checks (for influential observations) are performed.

| Table 3: Concise Set of Regressors |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Pop \& Cov Wgt | cov>. 6 | cov>. 75 | $\begin{gathered} \text { cov >. } 6 \\ \text { pop>. } 6 \mathrm{M} \end{gathered}$ | $\begin{aligned} & \text { cov >.75 } \\ & \text { pop>.6M } \end{aligned}$ | Avg |
| Daily Per Capita Music Radio (Hours) | -0.8091 | -1.2560 | -1.5237 | -0.6347 | -0.6931 | -0.9833 |
|  | (0.065) | (0.069) | (0.101) | (0.033) | $\underline{(0.019)}$ | (0.057) |
| Average Household Income ( 000 s ) | 0.0177 | 0.0194 | 0.0347 | 0.0084 | 0.0201 | 0.0200 |
|  | (0.033) | (0.079) | (0.044) | (0.320) | (0.009) | (0.097) |
| 2003 Internet Access | -2.1177 | -2.9273 | -4.2516 | -2.4070 | -2.2478 | -2.7903 |
|  | (0.026) | (0.053) | (0.073) | (0.005) | (0.018) | (0.035) |
| Observations | 95 | 61 | 41 | 53 | 36 |  |
| R-squared | 0.076 | 0.074 | 0.137 | 0.147 | 0.284 |  |
| Robust Regressions |  |  |  |  |  |  |
| Daily Per Capita Music Radio (Hours) |  | -0.7562 | -0.7493 | -0.7066 | -0.6614 | -0.7184 |
|  |  | (0.019) | (0.035) | (0.028) | (0.055) | (0.034) |
| Average Household Income ( 000 s ) |  | 0.0128 | 0.0146 | 0.0142 | 0.0187 | 0.0150 |
|  |  | (0.065) | (0.079) | (0.047) | (0.024) | (0.054) |
| 2003 Internet Access |  | -1.9139 | -1.7411 | -2.1668 | -2.0606 | -1.9706 |
|  |  | (0.009) | (0.043) | (0.003) | (0.015) | (0.018) |
| Observations |  | 61 | 41 | 53 | 36 |  |
| R-squared |  | 0.163 | 0.169 | 0.205 | 0.232 |  |

Robust p values in parentheses; p value for music radio is for one tail test; bold is sig at $10 \%$ level; bold underlined at $5 \%$, bold double underline $1 \%$; Constant term not shown.

Due to the relatively small number of observations it is important to try to maximize the efficiency of the estimates. To this end the regressions were rerun using only the variables that appear to actually have consistent and significant impacts-music radio use, Internet use, and income. The results are found in Table 3. The top half of that table provides the first differenced OLS regression
coefficients. The general results are similar but generally stronger than in Table 2. The coefficient on music radio is somewhat larger, averaging -.98 with a $p$ value in the range of .06 . The bottom half of the table provides the results from the robust regressions. When the impact of influential observations is weakened, the music radio coefficient is about the same as with the full set of regressors, -.72 , but more precisely measured with a $p$ value around $.03 .^{18}$

## B. The Nature of the Substitution

We have found that increases in time spent listening to music radio does not increase the purchase of sound recordings and it is more likely that radio play decreases the sale of sound recordings by an economically large amount. There are two possible explanations for a negative impact. One explanation might be that the time spent listening to radio is time that is taken away from other general entertainment activities and that listening to sound recordings is just one of these activities. The other explanation, which is the one that has been put forward in this paper, is that listening to music radio is a substitute for non-specific music listening using sound recordings.

Fortunately, it is fairly easy to test between these two possibilities. Not only do we have a measure of time spent listening to music radio but we also have a measurement of the time spent listening to talk radio. If the former hypothesis were true, talk radio would have the same impact on record sales as does music radio since time would be the key element of substitution and an hour of talk radio takes as much time as an hour of music radio. If the latter hypothesis were true music radio would have a more powerfully negative impact on sound recording sales than would talk radio.

[^8]| Table 4: First Differences Regression |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  <br> Cov Wgt | cov>.6 | cov>.75 | cov >.6 <br> pop>.6M | cov $>.75$ <br> pop>.6M |
| Daily Per Capita Music | -0.6238 | -1.1435 | -0.4070 | -0.8487 | -0.6004 |
| Radio (Hours) | $(0.126)$ | $\mathbf{( 0 . 0 8 2})$ | $(0.364)$ | $\mathbf{( 0 . 0 1 7 )}$ | $(0.113)$ |
| Daily Per Capita Talk <br> Radio (Hours) | 0.3996 | 0.2398 | 1.9753 | -0.5094 | 0.1904 |
|  | $(0.598)$ | $(0.842)$ | $(0.212)$ | $(0.319)$ | $(0.735)$ |
| Observations | 95 | 61 | 41 | 53 | 36 |
| R-squared | 0.08 | 0.08 | 0.17 | 0.16 | 0.29 |
| Test for Equality of <br> coefficients (p-val) | $(0.182)$ | $(0.266)$ | $(0.120)$ | $(0.437)$ | $\mathbf{( 0 . 0 3 6 )}$ |
|  |  |  |  |  |  |

Robust p values in parentheses; p value for music radio is for one tail test; bold is sig at $10 \%$ level; bold underlined at $5 \%$, bold double underline $1 \%$

Table 4 presents the (partial) results of regressions which include both talk and music radio in regressions otherwise identical to Table 3 (the concise set of regressors). The coefficient on talk radio is very imprecisely measured and generally positive. The coefficient on music radio is now somewhat smaller (in absolute magnitude) -.72. Certainly, talk radio does not appear to have the same impact or sign as music radio.

Because talk radio is so imprecisely measured we can only reject equivalence of the two coefficients for one specification; the other specifications have p-values ranging from .12 to .44 .

Nevertheless, talk radio certainly appears to be different than music radio. Our conclusion , therefore, is that music radio is a direct substitute for sound recordings independent of the time taken listening to radio. This is really not much of a surprise.

## V. Further Checks

## A. Outliers and Cutoffs

One possible issue is the impact of outliers. I examined all the results using the robust regression technique built into Stata beyond those instances mentioned in the text and the results were in close
agreement with those presented in the text. I also examined the DfBetas for the radio coefficient and concluded that the results presented are not due to a small number of influential observations.

It is also possible that the cutoff points chosen may have inadvertently impacted the results relative to other possible cutoff values. Examining other cutoff values (based on the concise regression specification), as shown in Table 5, reveals that the cutoff values chosen did not lead to unusual results. [Note that as some cutoffs change the observations may not change.] Interestingly, as the coverage cutoff increases, the coefficient for music radio rises but as the population cutoff rises the coefficient falls. An examination of p-values, also reveals that the chosen cutoff points chosen do not provide unusual results.

| Table 5: Music Radio Coefficients (and p-values) for Different Cutoff Values |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| PoplCov | 0.5 | 0.550 | 0.6 | 0.650 | 0.7 | 0.750 | 0.8 | Average |
| none | -0.9572 | -1.1719 | -1.2560 | -1.4040 | -1.4725 | -1.5237 | -2.0500 | -1.4050 |
| 400 | -0.9289 | -1.0739 | -1.1722 | -1.4040 | -1.4725 | -1.5237 | -2.0500 | -1.3750 |
| 500 | -0.9517 | -1.0974 | -1.2012 | -1.4414 | -1.5140 | -1.6070 | -2.1453 | -1.4226 |
| 600 | -0.4671 | -0.5597 | -0.6347 | -0.8320 | -0.7323 | -0.6931 | -0.6219 | -0.6487 |
| 700 | -0.4632 | -0.5597 | -0.6347 | -0.8320 | -0.7323 | -0.6931 | -0.6219 | -0.6481 |
| 800 | -0.3684 | -0.4496 | -0.5162 | -0.6963 | -0.6993 | -0.6296 | -0.5314 | -0.5558 |
| Average | -0.6894 | -0.8187 | -0.9025 | -1.1016 | -1.1038 | -1.1117 | -1.3368 | -1.00921 |
| p values |  |  |  |  |  |  |  |  |
| PoplCov | 0.5 | 0.550 | 0.6 | 0.650 | 0.7 | 0.750 | 0.8 | Average |
| none | (0.0945) | (0.0805) | (0.0690) | (0.0555) | (0.1005) | (0.1005) | (0.0615) | (0.0803) |
| 400 | (0.1140) | (0.1020) | (0.0860) | (0.0555) | (0.1005) | (0.1005) | (0.0615) | (0.0886) |
| 500 | (0.1120) | (0.1020) | (0.0855) | (0.0550) | (0.1000) | (0.0955) | (0.0590) | (0.0870) |
| 600 | (0.0700) | (0.0530) | (0.0325) | (0.0050) | (0.0265) | (0.0185) | (0.0635) | (0.0384) |
| 700 | (0.0735) | (0.0530) | (0.0325) | (0.0050) | (0.0265) | (0.0185) | (0.0635) | (0.0389) |
| 800 | (0.1210) | (0.0965) | (0.0645) | (0.0125) | (0.0260) | (0.0240) | (0.0915) | (0.0623) |
| Average | (0.0975) | (0.0812) | (0.0617) | (0.0314) | (0.0633) | (0.0596) | (0.0668) | (0.0659) |

## B. Simultaneity

Finally, another potential problem with the estimation performed so far is the possibility of simultaneity. We have examined the role of radio broadcasts on the sales of sound recordings. The argument might be made that the sales of sound recordings have an impact on radio listening just as
radio has an impact on sound recording sales. After all, they are substitutes for each other when individuals want to listen to non-specific music. The question is: is the amount of time individuals spend listening to radio a function of the number of sound recordings that they purchase?

Although a linkage is clearly possible, there are reasons to doubt the importance of sound recording purchases on radio listening. First, the number of sound recordings that an individual can listen to is the stock of owned recordings which is much larger than the flow of purchases, so the current flow might be at most only weakly related to the number of purchases unless the stock of older CDs depreciates rapidly over time. Second, for specific music consumption, sound recordings are the much preferred solution and radio will not be much of a substitute. Sound recording purchases intended mainly for specific (as opposed to general) listening should not, therefore, impact time spent listening to radio. Thus, only the portion of sound recording purchases intended for general consumption are likely to impact time spent listening to radio and this is even further dwarfed by time spent with radio. Third, we have seen that time spent listening to sound recordings is dwarfed by time spent listening to radio, so the impact of sound recording purchases on radio use is likely to be far less important than the impact of radio on sound recordings.

It is also useful to consider factors that might change the number of sound recordings purchased and the impact on radio listening. One very important factor during this period is file-sharing, and to this we should add instances of non-Internet based sharing, such as ripping borrowed CD s. Although we have a variable for file-sharing, it is Internet based and might not pick up all of the impact. If it did not, individuals would decrease their purchase of sound recordings and at the same time likely decrease their listening to radio since they can now have a very large free library of music to which they can listen. In this case, a reduction in record sales would be associated with a decrease in radio listening, not an increase.

Further problems with instrumental variables in this case have to do with the fairly small sample size. Instrumental variables are biased and inefficient, so it is not clear that, even if good instruments could be found, that we would be better informed from the use of IV.

Nevertheless, we can perform a test to determine whether there is evidence of simultaneity or not. The test is a form of Hausman specification test in which we regress radio music listening on a set of exogenous variables, calculate the residuals, and then include those residuals in the regression of record sales. In this case the exogenous variables include all the demographic variables used in the above regression plus, for the regression on radio music listening, changes in both the number of radio stations and time spent listening to talk radio, each of which should be independent of the musicradio/ sound-recording choice. Table 6 reports the coefficients on the residual variable for our various combinations of cutoff, which are insignificant with all cutoff values.

| Table 6: Coefficients of Residuals in Hausman Test |  |  |  |  |  |
| ---: | :---: | :---: | :---: | :---: | :---: |
|  |  <br> Cov Wgt | cov>.6 | cov>.75 | cov >.6 <br> pop>.6M | cov >.75 <br> pop>.6 <br> M |
| coefficient | 0.2619 | 0.1648 | 1.0382 | -0.7221 | 0.2765 |
| p value | $(0.812)$ | $(0.928)$ | $(0.654)$ | $(0.539)$ | $(0.784)$ |

The conclusion that would be drawn from this is that there is no simultaneity problem to worry about. Nevertheless, this test cannot be considered conclusive so we proceed to use instrumental variables in order to fully expunge the possibility of simultaneity.

Equation (1) represents the equation that we have been estimating with OLS up to this point. Equation (2) represents a structural equation explaining music radio usage. The two new variables in this equation are the number of radio stations (Stations) and the amount of time that individuals spend listening to talk radio (RadioTalk).
(1) Albums $=\mathrm{a}_{0}+\mathrm{a}_{1}$ RadM $+\mathrm{a}_{2}$ Inc $+\mathrm{a}_{3} \mathrm{BA}+\mathrm{a}_{4} \mathrm{Yng}+\mathrm{a}_{5}$ Male $+\mathrm{a}_{6}$ Old $+\mathrm{a}_{7}$ Int $+\mathrm{a}_{8}$ Minority $+\mathrm{a}_{9}$ Pop
(2) RadM $=b_{0}+b_{1}$ Albums $+b_{2}$ Stations $+b_{3}$ RadioTalk

Listening to talk radio fulfills a very different taste than does listening to sound recordings and should not be a substitute for listening to sound recordings, at least no more than any other activity that takes up time. Further, we have already seen that the time spent listening to talk radio does not impact the number of albums sold. If talk radio is independent of album sales, it should be uncorrelated with the error term in the regression on Albums.

Our other instrument is the number of stations in a DMA, which is a construct based on the average number of stations found in Arbitron metro areas weighted by the populations of the metro areas and as such doesn't relate directly to any particular set of physical stations since a single station can appear in more than one metro area. ${ }^{19}$ We expect this count of stations to be independent of record sales except through its impact on the radio music-use variable. The number of stations is determined in part by regulations since radio stations need government permission to broadcast. The number of stations may impact the variety of programming and might allow listeners to find programming closer to their tastes, impacting the time spent listening to music radio, but there does not appear to be any other way in which the number of stations should impact the sales of albums.

Our procedure will be to instrument for RadM in equation (1) with the fitted values of RadM from equation (3) that includes all the other exogenous variables that are found in equation (1) and the two instruments where $\mathrm{X} 1 \ldots \mathrm{X} 9$ is a vector representing variables 2-10 in equation (1).

$$
\text { (3) RadM }=\mathrm{c}_{0}+\left|c_{1} \ldots \ldots . c_{9}\right|\left|\begin{array}{l}
X 1 \\
\cdot \\
\cdot \\
X 9
\end{array}\right|+\mathrm{c}_{10} \text { Stations }+\mathrm{c}_{11} \text { RadioTalk }
$$

[^9]The results of the second stage regression coefficients for radio music are found in Table 7. As a byproduct of using instrumental variables, the coefficient on radio music is measured less precisely than is the case for OLS and the coefficient bounces around considerably more and in one instance it is even positive. Nevertheless, the average coefficient is about the same as before ( -.826 ) which further supports the view that there is no evidence that the OLS estimates are impacted by simultaneity.

The Sargan test for instrumental validity implies that our instruments are likely to be valid and not related to the error term. The Hansen J Statistic, which differs from Sargan in that it is robust in the face of heteroskedasticity, provides a less sanguine answer to the same question although it too suggests, but more weakly, that the instruments are valid. A test similar to the simultaneity test reported in Table 6 but robust to heteroskedasticity leads to the same conclusion as before- there is no evidence that music radio is endogenous and thus no need for instrumental variables to begin with. Finally, the Anderson canonical correlation likelihood ratio test tells us that the instruments identify the equation. The bottom of Table 7 provides some coefficients and other results from the first stage regressions where it is easy to see that the two variables used as instruments are highly correlated with changes in music radio usage.

| Table 7: Second stage IV estimates of change in sound recording sales |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | Pop \& Cov Wgt | $\begin{array}{c\|} \hline \text { Coverage } \\ >.60 \\ \hline \end{array}$ | $\begin{gathered} \hline \text { Coverage } \\ >.75 \\ \hline \end{gathered}$ | $\begin{aligned} & \mathrm{Cov}>.6 ; \\ & \text { pop>.6M } \end{aligned}$ | $\begin{gathered} \hline \mathrm{Cov}>.75 ; \\ \text { pop>.6M } \\ \hline \end{gathered}$ |
| radio music change in hours* | -0.9375 | -0.9658 | -2.2727 | 0.4015 | -0.6441 |
| p values (one tail) | (0.177) | (0.256) | (0.021) | (0.301) | (0.100) |
| Sargan [non heteroskedasticrobust] Instrument validity; Pvalue | 0.4303 | 0.4193 | 0.4112 | 0.7297 | 0.659 |
| Hansen J Statistic on instrument validity [hetero robust Sargan]; P-val | 0.2178 | 0.1278 | 0.1044 | 0.5537 | 0.4436 |
| Heteroskedastic robust [quasiHausman] exogeneity test; Chisq p value for RadM | 0.931 | 0.7223 | 0.5618 | 0.2698 | 0.9549 |
| Anderson Canon Corr Underidentication LR test; p value | 0.0000 | 0.0000 | 0.0001 | 0.0008 | 0.0004 |
| Partial First Stage Results; Music Radio is dependent variable |  |  |  |  |  |
| station count change | 0.0025 | 0.0025 | 0.0023 | 0.0022 | 0.0024 |
|  | $\underline{0.000}$ | $\underline{0.000}$ | (0.001) | (0.004) | (0.002) |
| radtalkchg | -0.6657 | -0.5962 | -0.6768 | -0.5600 | -0.6778 |
|  | $\underline{0.000}$ | $\underline{0.000}$ | $\underline{0.000}$ | $\underline{\mathbf{0 . 0 0 0}}$ | $\underline{\mathbf{0 . 0 0 0}}$ |
| Observations | 90 | 61 | 41 | 53 | 36 |
| R-squared | 0.537 | 0.486 | 0.642 | 0.476 | 0.626 |
| Robust p values in parentheses; *=instrumented variable; bold is sig at $10 \%$ level; bold underlined at $5 \%$, bold double underline $1 \%$ |  |  |  |  |  |

We conclude that simultaneity is not a problem for our OLS results.

## C. Errors in Variables

Although we have taken steps in our estimation to eliminate or weaken any impact of measurement error, one might argue that such errors cannot have been completely eliminated. It is well known that under classical errors-in-variables situations (which assumes the measurement error term is not correlated with the true values of the variables) coefficients on all the rhs variables will be biased and inconsistent if any of the variables is mismeasured.

Of course, our interest is centered on the coefficient for music radio listening. If there were only one explanatory variable in the regression the nature of the bias due to the mismeasurement is much
easier to determine since it would simply become the error-in-variables attenuation bias, where the coefficients are biased toward zero. For this reason the regressions were rerun leaving out all the rhs variables except music radio listening time. Table 8 shows that the results from these regressions are very similar to those obtained from the complete regression. Under the standard assumption, therefore, we can conclude that measurement errors are likely to lower our estimates of the impact of music radio.

| Table 8: Regression with Radio Music Use as Sole Independent Variable |  |  |  |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  <br> Cov Wgt | Coverage <br> $>.6$ | Coverage <br> $>.75$ | Cov $>.6 ;$ <br> pop>.6M | Cov $>.75 ;$ <br> pop>.6M | Average |
| sole variable | -0.7505 | -1.0323 | -1.1118 | -0.3877 | -0.4976 | -0.7560 |
| p value | $(0.113)$ | $(0.143)$ | $(0.200)$ | $(0.157)$ | $\mathbf{( 0 . 0 7 3 )}$ | -0.1369 |
| Observations | 95 | 61 | 41 | 53 | 36 |  |
| R-squared | 0.024 | 0.024 | 0.031 | 0.013 | 0.043 |  |
| In Full regression | -0.7903 | -0.7507 | -1.1817 | -0.6049 | -0.7767 | -0.8209 |
| From Table 3 | $\mathbf{( 0 . 0 7 6 )}$ | $(0.169)$ | $(0.126)$ | $\mathbf{( 0 . 0 6 7 )}$ | $\mathbf{( 0 . 0 5 6 )}$ | -0.0985 |
| R |  |  |  |  |  |  |

Robust $p$ values in parentheses; $p$ value for music radio is for one tail test; bold is sig at $10 \%$ level; bold underlined at $5 \%$, bold double underline $1 \%$

Knowing that the true coefficient is larger than the measured coefficient is probably sufficient for our purposes, since it merely strengthens the conclusions already drawn. A solution often proposed for errors-in-variables is to use instrumental variables. Although we have performed such as examination above, there are difficulties with using it as a salve for the errors-in-variables problem beyond the difficulties mentioned for issues of simultaneity. Among those difficulties is the fact that most potential instruments (including the ones chosen) will suffer from the same errors-in-variable problems as the OLS results unless instruments could be found at the DMA level as opposed to the MSA level.

## VI. Gauging the Overall Impact of Radio

We have found that radio use lowers sales of sound recordings. But we have only a limited range of observations to work with. Yet the regression results that we have found are compatible with other scenarios. For example, radio at first might have a positive informational aspect on sales, which then turns negative when greater radio use becomes a substitute for listening to CD s, as discussed earlier. In
this case the overall impact of radio could be positive or negative in spite of our findings. Assume, for the sake of example, that radio has a positive impact for approximately the first .5 hours of daily use and a negative impact thereafter. This is illustrated in Figure 1 for three possible cases, A, B, and C.

Correctly estimating the impact of music radio when all observations are between 1.5 and 3 will lead to a conclusion that music radio lowers record sales, which is correct within the bounds of our examination. Attempting to extrapolate the impact of a factor, such as radio use, to levels that are outside the bounds of the sample can easily provide misleading results if the relationship looks like A or B, however. The negative relationship found in A would obscure the overall positive impact that radio play had on sound recordings since the large positive impact from the first half hour of music radio would be obscured.

Figure 1: Out of Sample Estimates can be Misleading


The bounds of music radio use in our 2003 sample run from a low of 1.9 hours to a high of 3 hours, with an average of 2.3 hours. The 1998 values are just slightly higher. The range of changes in music radio use is .6 hours from 1998 to 2003. Within these ranges of observations the measured impact of radio play on the sales of sound recordings is negative. The average album consumption stood at 2.3 units per capita in 2003. If we were to assume that the relationship between music radio and CD purchases were linear throughout its range, as illustrated in case $C$, an increase in radio use from 0 to 2.3 hours per day could be expected to reduce album sales by more than one and a half albums, given a coefficients of -.75 . This would be a large negative impact of overall radio use. Yet the relationship represented by curve B would imply a loss of only 1 unit and the relationship represented by A would imply a gain of 1 unit.

Is there any evidence for or against such a nonlinearity that might overturn the results found in these regressions? First, we ran quadratic specification of the amount of radio music use to see whether there was any evidence of nonlinearity within our data. There was not. We also split the data in half based upon music radio usage and ran separate regressions for each half. The cities with smaller music radio usage had a larger negative impact than the cities with greater music radio usage, contrary to what we would expect from the type of nonlinearity suggested by lines A or B. Still, the limitations on our data keep us from being able to rule out this type of possible non-linearity.

The historical approach used in Liebowitz (2004), however, can be used to throw some light on this possible nonlinearity. In that paper I examined the sales of sound recordings immediately before, after, and during the introduction of radio into the American market. If there was an overall positive promotional element in radio, and if it were large, it should have shown up in those statistics since the initial impacts of the very first hour or two of music listening per week would be included in the statistics, which measured the total impact and not marginal impacts. As already mentioned, the sound recording market was already quite mature at that time with per capita sales the equivalent of those in
1950. Y et, as that paper reported, there was no evidence of any but a negative impact of radio on sound records since sales fell significantly during the first few years of radio's growth in spite of a healthy and growing economy. The fact that record sales fell during the birth of radio would seem to imply that the net effect is negative, even at an initial stage.

A general historical examination of the sound recording industry (Morton 2003) reached a similar conclusion:

Record companies welcomed the subsequent transfer of electrical technology from radio and motion pictures to the phonograph industry, but hated the effect these two new forms of entertainment had on the record business. Radio was the biggest threat. On the eve of broadcasting's debut, between 1914 and 1921, record sales had doubled, largely because of sales of popular music. With the inauguration of network radio in the middle 1920s, the market for popular recordings collapsed, resulting in a number of companies leaving the field or changing ownership. (Page 26).

Even if the net effect quickly goes negative, that doesn't mean that there are no initial positive impacts but only that such positive effects are quickly overwhelmed by the negative effects. To be sure, this issue cannot be completely settled since one can argue the radio/ sound-recording relationship in the early 1920s might have been very different than the current relationship. We cannot know how correct this claim might be. Nevertheless, the lack of any evidence in favor of the possibility of a net positive impact, when compared to the much more plentiful evidence of the negative impact of music radio, would seem to provide a prudent analyst with the less than perfectly held conclusion that radio has a net negative impact on sound recording sales.

## VII. Discussion

The existence of payola, the now illegal payment made by record companies to radio station personnel, could be taken as evidence that radio stations generate sufficient positive impact on record
sales that the typical market clearing price for the right to broadcast sound recordings would be negative price for the rights to a sound recording.

Both Richard Caves (2000) and Ronald Coase (1979) have written on the history of payola. They both note that numerous attempts were made by record companies and before them, music publishers, to stop paying radio station personnel or well-known performers, beginning, according to Coase, with an episode in 1890. Some of these attempts, such as the congressional hearings in the late 1950s, appear to be instances where established record companies were trying to reduce the airplay of a group of smaller record companies who were heavy users of payola and who happened to specialize in that evil music, rock-and-roll. Caves also suggests that modern attempts to limit payola also favor major record companies over independents.

There may well be truth to these claims of redistributional impacts from attempts to control payola. Nevertheless, if radio play benefited radio stations in an overall sense, the industry should not have wanted to eliminate the practice altogether, unless the amount of music played was going to be independent of such payments.

In order to reconcile the results in this paper with the existence of payola note that radio play of music, in an overall sense, was largely out of the control of the record companies. There does not appear to be any evidence, for example, that record companies try to convince radio stations to increase the share of music relative to talk, or to try to convert talk radio stations into music radio stations.

An overall negative impact of radio play is consistent with the concept of payola for the simple reason that payola doesn't impact the total quantity of radio broadcasts of sound recordings. Payola only impacts which particular songs are broadcast. The overall negative impact of radio play measured here, would be beyond the feasible control of record companies due to the current lack of broadcast property rights in sound recordings. Any record company that attempted to, let's say, pay radio stations
to play fewer hours of sound recordings would only receive a portion of the benefits which would accrue to all sound recording companies. Antitrust laws would prevent the entire industry from making such payments. Further, any station (talk radio, say) could then threaten to play more sound recordings (by changing formats) in order to extort additional payments

The results of this paper are entirely consistent with a modified version of the suggestions of these two esteemed economists. They both propose that payola should not be illegal, that it is payment for a useful service, and that the marketplace should be used to determine what payments should be.

For example, Coase concludes (p 318):


#### Abstract

..if the playing of a record by a radio station increases the sales of that record, it is both natural and desirable that there should be a charge for this. If this is not done by the station and payola is not allowed, it is inevitable that more resources will be employed in the production and distribution of records, without any gain to consumers, with the result that the real income of the community will tend to decline. In addition, the prohibition of payola may result in worse record programs, will tend to lessen competition, and will involve additional expenditures for regulation.


Caves states (p 292):

The evidence supports a simple interpretation of the economics of payola in broadcasting. Promotional benefits to the label cannot be captured directly by the broadcaster, who lives by advertising revenue that generally will not reflect this benefit. Payola compensates for valuable promotion, and leaves us wondering why it is stigmatized as bribery rather than recognized as payment for services rendered.

We agree completely with their call for a fully functioning market. A complete market, however, would not merely allow payola to be legal. A fully functioning market would allow a complete set of property rights over the sound recording being broadcast, including the ability to restrict radio play and to provide geographic exclusives.

## VIII. Conclusions

The impact of radio broadcast of music on the sales of sound recordings has received scant attention by researchers. Y et the economic relationship between these two industries is likely to become increasingly important in the near future as the transmission of music becomes increasingly digitized and the property rights of the copyright owners come under greater scrutiny and political pressure.

The analysis above indicates that radio play is largely a displacement for the sales of sound recordings. The panel data provide evidence that radio play is negatively related to the overall level or record sales within major cities. Although the levels of statistical significance only bordered on typical levels of statistical significance, the similarity of the overall results for different subsets of observations and for different specifications provide an additional level of support for this conclusion.

The negative impact of radio only exists for music broadcasts and not talk radio, which is consistent with a view that listening to music on the radio is a close substitute for listening to music on sound recordings. The measured negative impact of music radio on record sales is in the vicinity of $20 \%$ within the range of our observations. Extrapolating these results outside the bounds of our sample provides for a considerably larger impact, although such extrapolation is fraught with difficulties. Those difficulties are ameliorated somewhat by appealing to other evidence and other tests.

On a methodological note, the apparent divergence between the impact of radio play for individual records versus that for the entire industry indicates an important danger in trying to estimate the impact of a factor such as radio play by examining the impact on individual recordings. This potential fallacy of composition should be kept in mind whenever there are reasons to believe that the behavior of the whole may be different than the behavior of the individual parts (besides radio broadcasting, the example of file-sharing's impact on individual recordings vis-à-vis the entire recording
industry come to mind). In this case, alterations in share occurs independent of changes in the overall market and it is important not to conflate share changes with market changes.

These results also provide some suggestions for public policy that is likely to become increasingly important in the next few years as new broadcasting techniques to make using the radio an even closer substitute for the purchase of sound recordings.

With a full property rights system in place, record companies could control how frequently their records were played and extract payments from radio broadcasters or make payments to broadcasters, as the case might be. A complete market solution would look like the one between television and movies. Record companies should be able to enter into whatever contracts they wish, including restricting the playing of songs to particular stations in particular localities. With this additional proviso, the market solution suggested by both Coase and Caves can be readily supported. In that case the true impact of radio play can be determined where it is best determined-in the market.

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## IX. Cross Section Results by Year

Because the simple cross section results are likely to be eclipsed in usefulness by the fixed effects results we will not dwell on them. Table 2 presents results from for 1998 and 2003. Our primary interest is in the relationship between time spent listening to music radio and the purchase of sound recordings. Because the generally accepted hypothesis is that radio play increases record sales, we make that our null and for that reason use a one tailed test of significance. The relationship appears strongly negative in 2003 but much more weakly negative in 1998.

|  | ---------1998---------- |  |  | ---------2003--------- |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Pop\&Cov | Coverage> .6 | Coverage $>.75$ | Pop\&Cov | Coverage $>.6$ | Coverage .75 |
| Daily Per Capita Music Radio (Hours) | -0.2684 | -0.3407 | -0.2231 | -0.8985 | -0.8684 | -0.7406 |
|  | (0.162) | (0.164) | (0.300) | (0.004) | (0.012) | (0.060) |
| Average Household Income (000s) | 0.0014 | -0.0037 | -0.0144 | 0.0038 | -0.0035 | 0.0009 |
|  | (0.905) | (0.825) | (0.420) | (0.620) | (0.705) | (0.942) |
| Internet Access | 2.8033 | 3.5014 | 3.7365 | 2.2326 | 3.2354 | 1.1920 |
|  | (0.003) | (0.004) | (0.008) | (0.013) | $\underline{(0.005)}$ | (0.325) |
| BA Degree or above | 2.0535 | 1.3688 | 3.8211 | 1.4250 | 1.3102 | 3.1834 |
|  | (0.153) | (0.495) | (0.076) | (0.280) | (0.332) | (0.190) |
| Share 12-29 | -1.4090 | -1.8482 | 1.3650 | -5.9985 | -6.6625 | -6.6705 |
|  | (0.319) | (0.354) | (0.481) | (0.004) | (0.001) | (0.117) |
| Share Males | -0.0535 | -0.4676 | 0.6412 | -2.5706 | -5.7907 | -6.3801 |
|  | (0.976) | (0.842) | (0.796) | (0.501) | (0.268) | (0.248) |
| Share 55+ | -2.3272 | -2.0592 | -1.5963 | -2.8457 | -4.0333 | -6.0944 |
|  | (0.063) | (0.224) | (0.397) | (0.165) | (0.115) | (0.118) |
| Share Minority | -0.1631 | -0.1207 | -0.0721 | 1.5137 | 1.4869 | 1.2157 |
|  | (0.705) | (0.831) | (0.902) | (0.002) | (0.011) | (0.069) |
| DMA Population | 0.0025 | 0.0023 | 0.0019 | -0.0003 | -0.0006 | -0.0018 |
|  | (0.023) | (0.077) | (0.098) | (0.776) | (0.661) | (0.195) |
| Constant | 3.1779 | 3.7826 | 1.7483 | 5.9602 | 7.6909 | 8.9437 |
|  | (0.032) | (0.080) | (0.377) | (0.086) | (0.088) | (0.123) |
| Observations | 94 | 62 | 42 | 92 | 66 | 47 |
| R-squared | 0.505 | 0.491 | 0.669 | 0.53 | 0.5 | 0.529 |

Robust $p$ values in parentheses; $p$ value for music radio is for one tail test; bold is sig at $10 \%$ level; bold underlined at $5 \%$, bold double underline $1 \%$

Cities with populations having greater financial resources and media expertise would be expected to purchase more sound recording albums. Income, possession of the college degree and Internet

Access all measure some dimension of this characteristic and are highly correlated with one another (~.6), although the Internet Access variable is related to file-sharing in 2003, as discussed in more detail below. Although the coefficients for these three variables are generally consistent with this hypothesis, the variables based on the share of individuals with Internet access and BA degrees are far more influential and significant than is household income.

Cities with larger shares of males and youthful individuals have lower record sales in 2003 but not in 1998, although file-sharing might be responsible for much of this since both groups are much more likely to engage in file sharing. Minorities are associated with higher record sales in 2003, but there is no impact in 1998. Larger cities seem to be associated with greater record sales in 1998, but there is no impact in 2003. An increased share of individuals over 55 appears to decrease record sales which makes sense since older individuals do not purchase many records according to RIAA surveys. Minority status appears positively related to record sales in 2003, but not 1998 and this difference is puzzling if not and perhaps somewhat troubling.

These cross section results should be taken with a certain amount of caution. It is generally understood that cross section results are often less reliable than similar panel data since panel data allow the control of fixed effects that might not be picked up in the cross section regressions. For example, there may be important differences between cities that we are not controlling for, such as the role of music in everyday life, technological and media knowledge, the importance of ethical or religious beliefs, immigration patterns, or family structure. If the left-out factors do not vary over time (major socioeconomic factors are unlikely to greatly change in five years) then the use of panel data will be preferred to individual cross section regressions. We now turn to such an approach.


[^0]:    ${ }^{1}$ The sole exception of which I am aware is Liebowitz (2004).
    ${ }_{3}^{2}$ Sometimes payola but more frequently intermediaries who can bypass the laws against payola.**
    ${ }^{3}$ In contrast to record companies, movie owners are able to strictly control whether, when, for what price, and who gets to broadcast their movies. Selling geographic exclusivity in these rights to single stations is common. Analyzing the historical reasons for this different set of rights granted to movie owners versus sound recording owners is beyond the scope of this paper, but there are several possibilities: 1) there was no copyright on sound recordings until 1971 so there was no right to sell and inertia might have kept it this way; 2) the belief that radio was beneficial to sound recording sales implied a zero or negative price; or 3) sound recording firms had less political power vis-à-vis radio broadcasters than did movie owners relative to television broadcasters and thus the sound recording owners were unable to secure for themselves the same set of rights as movie producers.

[^1]:    ${ }^{4}$ In some countries owners of sound recordings receive a form of payment for the use of their recordings (with rates usually set by law or supervised by some quasi-judicial organization) although owners of sound recordings cannot restrict usage and must make their recordings available to any radio station making the required payment. There is also ${ }^{5}$ "performance right" for the underlying musical compositions, paid to composers and their publishers.
    ${ }^{5}$ See for example Washington Post, "Music Labels Sue XM Over Recording Device" Annys Shin, May 17, 2006; Page D01 at http://www.washingtonpost.com/wp-dyn/content/article/2006/05/16/AR2006051601826.html

[^2]:    ${ }^{6}$ Reported in Figure 1 in Thomas Hazlett, "Physical Scarcity, Rent Seeking, and the First Amendment" Columbia Law Review, Vol. 97: 905-944. Hazlett's data are taken from Bureau of the Census.
    ${ }^{7}$ See Liebowitz (2004).
    ${ }^{8}$ See http://www.garlic.com/~tgracyk/tmw.htm.
    ${ }^{9}$ This number comes from correspondence with the Recording Industry Association of America (RIAA) as reported in Liebowitz (2004).

[^3]:    ${ }^{10}$ Radio (including satellite) is listed at 2.75 hours per day and sound recordings at .5 hours per day. See Table 1116 "Media Usage and Consumer Spending for 2003." The ratio was closer to $3: 1$ in 1999, before file-sharing began. Available at http://www.census.gov/compendia/statab/tables/06s1116.xls.

[^4]:    ${ }^{11}$ In 1998 the radio genres which we classified as 'talk' were: News, Religion, Sports and Talk. In 2003 the genres had multiplied and changed, and we classified as talk: All News, All Sports, Educational, News Talk Information, Spanish News/Talk, Sports, Talk/Personality, and Religious. Note that Gospel, although religious, is classified as music.
    ${ }^{12}$ The control for file-sharing requires that the start date occur prior to file-sharing (1999) and that only one other year be used. For details see Liebowitz (2006).
    ${ }^{13}$ The Census Data also include PMSAs (primary metropolitan statistical areas) and CMSAs (consolidated metropolitan statistical areas) which are entire or parts of more heavily populated MSAs.

[^5]:    ${ }^{14}$ Arbitron states: "Arbitron Metros generally correspond to the Metropolitan Statistical Areas (MSAs, PMSAs, CMSAs) defined by the U.S. Government's Office of Management and Budget. They are subject to exceptions dictated by historical industry usage and other marketing considerations as determined by Arbitron." See page 8.2 of Arbitron Radio Market Report Reference Guide, 2002.
    ${ }^{15}$ Coverage ratios were calculated for each DMA for both Arbitron and Census data and the lowest ratio for either Arbitron or Census data is used for each DMA. One difficulty in constructing these ratios was that Nielsen populations were based on individuals over the age of 2 whereas Arbitron populations were based on individuals over the age of 12 . This required that we used Arbitron listed DMA populations when calculating the Arbitron coverage ratios.

[^6]:    ${ }^{16}$ The weighting was constructed taking the product of the squared coverage and the square root of the population.

[^7]:    ${ }^{17}$ I used the built in RREG Stata routine to determine whether weakening the impact of influential observations would change the results. Although the coefficients were slightly lower, the $p$ values were slightly stronger (. 08 versus .10 ). The RREG routine first eliminates observations with levels of Cook's D that are above 1 and then it iteratively lowers the weight for observations with large absolute residuals until a convergence threshold is reached.

[^8]:    ${ }^{18}$ Although the robust regressions were not shown for Table 2 , the average coefficient was .684 and the average $p$ value was . 079

[^9]:    ${ }^{19}$ Not all stations in a metro area were counted. If a station was listed as having an audience rating (percentage of audience) of zero, it was excluded from the analysis. This is similar to Arbitron's listings which include stations only if they have a measurable presence, although they do not base it on ratings points but instead on audience size.

