# Glass Ceilings in the Art Market 

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#### Abstract

Using a unique data set consisting of the population of fine art auctions from 2000 to 2017 for Western artists, with over 2.7 million auction sales, we provide strong empirical evidence for a glass ceiling for female artists. First, we show that female artists are less likely to transition from the primary (gallery) into the secondary (auction) market, where $96.1 \%$ of auction sales are by male artists. This higher bar materializes in an average price premium of $4.4 \%$ for artworks by female artists at auction. Second, this premium is driven by a small number of women located at the top of the market, and manifests as a discount of $10 \%$ when we account for the number of artworks sold per artist. Third, the superstar effect, where a small number of individuals absorb the majority of industry rewards, prevails; at the top $0.1 \%$ of the market, artworks by female artists are traded at a discount of $9 \%$. Moreover, the very top $0.03 \%$ of the market (where $40 \%$ of the sales value are concentrated) are currently unattainable for female artists selling at auction - revealing a glass ceiling. Our study has important implications for industries characterized by a superstar effect and illustrating how market structure impacts gender equality.


Keywords: Art market, Auctions, Gender economics, Labour economics
JEL Codes: J24, J31, J71, Z11

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## 1 Introduction

Market structures characterized by the superstar effect where a few individuals absorb the majority of rewards can reinforce the lack of mobility of underrepresented market participants. In these so-called "winner-take-all" industries lesser talent is an imperfect substitute for higher talent and the gap in compensation can be disproportional to the gap in skill (Rosen, 1981). Eventually, the size of rewards will depend more on rank than on talent. As stated by Frank and Cook (2013), in "winner-take-all" markets sellers who are not located in the top of the distribution often earn less than they could have earned in alternative occupations. At the same time, consumers in these markets are often not able to perfectly evaluate the level of skill making meritocracy more difficult to achieve. As a result, higher quality standards might be applied to market participants for whom fewer track records are available and for whom it was historically more difficult to acquire legitimacy due to lower status characteristics. ${ }^{1}$ Furthermore, information cascades and herding behavior where buyers imitate the decisions of preceding actors disregarding their own private information can result in market inertia (Banerjee, 1992; Bikhchandani et al., 1992). In summary, the extremely large competition for top ranks in combination with information asymmetries in these industries might inhibit underrepresented groups from progressing. This may lead to a glass ceiling for them as consumers try to minimize risk and avoid uncertainty in their consumption decisions.

In this large-scale empirical study, we are interested in whether the superstar effect encountered in the art market translates into barriers for female artists at auction. In particular, we investigate auction outcomes for artworks created by female and male artists and analyze whether the prevailing market structure impacts artwork prices and materializes in a glass ceiling for female artists at the top of the market.

The superstar effect is very well illustrated in the art market. The distribution of rewards in this industry is highly skewed with the largest profits concentrated on top. According to the latest Art Basel and UBS art market report (McAndrew, 2018) only $1 \%$ of artists accounted for $64 \%$ of auction sales in terms of value in 2017. Furthermore, the heterogeneous nature of art as a good makes it difficult to evaluate quality. These information asymmetries give a pivotal role to the institutions in the art world (for instance, dealers, auction houses and curators) in resolving uncertainties and in determining the superstars in the market. Since the birth of the modern art market in the mid-19th

[^1]century in London little has change in terms of the identity of the key instituions, sales channels as well as the structure of the competetive landscape (De Silva et al., 2017). It is also documented that female artists are historically underrepresented in this industry. The eminent female art historian, Linda Nochlin, was among the first ones to question the notion of the male genius and to draw attention to the issue of gender discrimination in the art world in her 1971 landmark essay Why Have There Been No Great Women Artists?. Later in 1984, the Guerrilla Girls started to create awareness for sexual discrimination by pointing out the underrepresentation of female artists in the New York based Museum of Modern Art's exhibition International Survey of Painting and Sculpture where only $10 \%$ of all works were by female artists (The Guerrilla Girls, 2017). The share of female artists decreases gradually along the career ladder and indicating impaired mobility. While women do not display less interest in pursuing an artistic career than men do (about $50 \%$ of all Master of Fine Arts (MFA) holders are female in the US), their share drops to $30 \%$ in commercial US galleries (National Museum of Women in the Arts, 2017) and to $25 \%$ at art fairs (McAndrew, 2018). According to the National Museum of Women in the Arts (2017), nowadays, artworks by female artists represent only $3 \%$ to $5 \%$ of major permanent collections in the US and Europe. Furthermore, based on our data, female artworks at auctions make up less than $4 \%$.

The auction market constitutes the secondary market for art where artworks are traded which were previously sold on the primary market (i.e. through galleries). This implies that these artworks have a resale value and are in demand by other buyers. The appearance of an artwork at auction signals professional recognition (Goetzmann et al., 2016). Therefore, artists who reach this stage are regarded as relatively established with their quality being certified by the market. Similar to individual wages these prices represent market value of an artist. An artist's track record within the secondary market is highly visible to the public as opposed to prices in the primary dealer market. Information on past auction results is frequently used by art collectors, experts and consultants as input to determine the value and future potential of an artist. Therefore, while prices fetched at auction do not directly accrue to the artists themselves they may have a large impact on their careers and can also feed back into gallery prices (Galenson and Weinberg, 2000).

Given the low representation of women in the art market, we expect the superstar effect to create larger barriers for female artists which will cause them to reach the secondary (auction) market less likely than male artists. However, conditional upon reaching this market, we predict that the
artworks by female artists should either command a price premium or show no significant difference compared to prices for artworks created by male artists after controlling for quality. The rationale for a premium for female artworks could be due to a stricter quality filter in earlier career stages which creates a higher bar for female artists. As a result, a female artist will have to be on average better than a male artist in order to transit from the primary into the secondary market. However, if being traded on the secondary market is not sufficient to signal quality (i.e. resolve uncertainty with respect to the ability and/ or the future potential of female artists) and establish legitimacy as an artist, a stricter quality standard might still be applied which should materialize in yet another hurdle along the career trajectory. In particular, we should see a glass ceiling arising at the top of the market meaning that female artists are less likely to reach the highest ranks than male artists.

We have exclusive access to a unique auction data set representing nearly the whole population of global art auction transactions in the time period between 2000 and 2017. From this global database we consider only Western artists to facilitate an accurate identification of the gender of the artists. Overall, we have auction results of 110,938 male and 5,612 female artists (with 2,572,346 and 104,844 lots respectively). Furthermore, the data covers several art movements and media allowing us to segment and homogenize sub-samples resulting in better comparability. This auction sample allows us to compare the performance of female artists in the secondary market to the performance of male artists. Our focus on the output (the artistic product) instead of on the input of labor as it is common practice in gender performance differential studies (Becker et al., 1957). This implies that our results less likely to be driven by differences in the individual characteristics of the artists. ${ }^{2}$ Additionally, we are also employing a unique primary (gallery) market data set to investigate the mobility between the primary and the secondary (auction) market for male and female artists. Lastly, we also perform a quantile regression analysis to test for the presence of a glass ceiling particularly in the upper tail of the price distribution where most of the rewards are concentrated.

Our results reveal strong evidence for the presence of structural barriers faced by female artists in the market for fine art. First, women are less likely to enter the secondary (auction) market. This selection bias may explain the average $4.4 \%$ price premium found for female artworks at auction. Second, this price premium is likely to be due to a supply squeeze caused by a small number of female artists located at the top $1 \%$ of the market. In particular, this effect can be attributed to

[^2]female artists from older generations (mainly pre-1950's works of art) where institutional barriers for women pursuing artistic career were stronger. At the same time, we observe an increase in the share of contemporary artworks by women traded at auction. These artworks sell at a price discount of $8.3 \%$ compared to contemporary artworks by men. This provides evidence of looser barriers for female artists in recent time periods. However, we also show that market concentration is higher for women than for men accentuating stronger competition for top ranks among female artists. This implies that the superstar effect, where rewards are concentrated among a few individuals, appears to be even more prevalent for women. Third, we identify a second hurdle faced by female artists which is manifested in their exclusion from the superstar league of the market. In particular, the top $0.03 \%$ of the market, where $40 \%$ of the industry rewards are concentrated, are off-limits for women and is entirely occupied by men. This is supported by the quantile regression results which reveal that artworks by female artists are traded at a price discount of $9 \%$ in 99.9 th quantile of the price distribution. Assuming that talent is evenly distributed across gender the exclusion of women from the superstar league of the market points to important structural barriers for underrepresented groups in the market for fine art.

We complement the literature on the influence of a market structure charactersized by the superstar effect on the performance of underepresented groups in particular within the area of cultural and creative industries. There is a lot of anecdotal evidence on gender inequality in the art world as well as surveys (Throsby and Mills, 1989; Throsby et al., 2003, 2010), case studies (Cowen, 1996; Reis, 1995a,b) and empirical studies (Adams et al., 2017; Cameron et al., 2017; Rengers, 2002). These document performance differences between male and female artists. Findings stress the high underrepresentaion, relatively lower sales revenues of female artists, as well as hurdles experienced by women that impede their careers. However, these study are too limited in scale and their rather small sample sizes impact the strength and robustness of the results. As superstar markets are characterized by a skewed distribution of income, it is of particular importance to be able to analyze the tails of these distributions which calls for a critical sample size. For instance, using a sample of 4,434 fine art graudates from Yale University from 1891 until 2014, a recent study by (Cameron et al., 2017) provides partial empirical evidence of a higher bar for women. They show that while female artists experience more obstacles than men when entering the market, their artworks sell at a premium conditional on being traded in the market. This effect is mainly driven by pre-1983 graduates.

The effect disappears for post-1983 graduates. While the authors use a rather small sample (auction results of 515 artists are examined) which is more likely to include higher skilled artists given the quality of the school, our results are consistent with these findings. Another recent larger empirical study by Adams et al. (2017) employs a sample size of 1.5 million global transaction over a time period from 1970 until 2013 studying performance difference between men and women at auction. The authors document an average discount of $47.6 \%$ for female artworks before adjusting for artist and artwork characteristics. This effect is moderated by the level of country-specific gender inequality. Our results might be different from their findings due to differences in the sample composition. Given that we have 2.7 million auction transactions over a period of 18 years for Western artists only our data is more dense representing almost the entire population of auction transactions for this sub-sample. There is some statistical support from other industries subject to superstar economics showing that women encounter a glass ceiling when they are climbing up the career ladder. This was found to be the case for top athletes (Kahn, 1991), movie stars (Bielby and Bielby, 1996; Lincoln and Allen, 2004), top executives (Bertrand and Hallock, 2001) and researchers (Barbezat and Hughes, 2005; Probert, 2005). However, these studies use salaries to investigate reward differentials which bears the risk of omitting individual characteristics such as negotiation skills that might explain the gap in salaries. Our focus on auction prices enables us to isolate the market side from the personal influence of the artist.

Our results provide important lessons on the impact of market structures that result due to the prevalence of the superstar effect. It appears that the competitive pressure for high ranks paired with uncertainty about the level of skill leads to higher quality standards for underrepresented market participants and impairs their mobility. While we cannot provide evidence of mechanism that causes the barriers that lead to the observed performance differences between male and female artists, we offer material input for a closer investigation of the underlying reasons. ${ }^{3}$ Our results can be extended to other occupational areas characterized by the superstar effect such as the high-end gastronomy, top sports, academia, journalism as well as for leadership positions in general which are still largely dominated by homogeneous groups.

[^3]The paper proceeds as follows: In section 2, we provide an overview of the current state of research with respect to gender performance discrepancies in the art world. The data set is described in section 3. Section 4 is dedicated to the empirical analysis and results. We finish with some concluding remarks in section 5 .

## 2 The Gender Gap in the Art Market

Due to a lack of data research on inequality in performance within the art market is mainly confined to surveys (Throsby and Mills, 1989; Throsby et al., 2003, 2010) and qualitative survey- or interviewbased case studies (Pheterson et al., 1971; Piirto, 1991; Cowen, 1996; Kirschenbaum and Reis, 1997; Brooks and Daniluk, 1998). In repeated surveys on the artist labor market in Australian, Throsby and Mills (1989); Throsby et al. $(2003,2010)$ found that the share of women has increased from $50 \%$ to $63 \%$ over the time period from 1983 until 2008. Further, they show that women feel more restrained in their work by having children. Reis (1995a,b) found in their studies that motherhood diverts time from creative endeavors and that a female artist's productivity peak is more likely to occur at a later age. The resulting income of female artists was found to be $38 \%$ below the income of male artists with a decreasing trend. In addition, women were found to be more often in the low income quantiles than men ( $44 \%$ and $35 \%$ ) and less often in higher income quantiles ( $16 \%$ and $25 \%$ ). However, these studies have very limited sample sizes and employ a broader definition of the artist profession by including performance artists and writers. For the case of US artists, Filer (1986) found that being female (married or unmarried) has a negative effect on earnings. Consistent with the findings above, Rengers (2002) reported that Dutch female visual artists sell their artworks for lower prices (about $30 \%$ ) and are less represented by agents. The author explains this with the fact that female artists are on average younger and have less experience than men. Lastly, a case study involving 10 female artists with children by Kirschenbaum and Reis (1997) found that creative productivity was related to access to financial resources, spousal support, childbearing responsibility and recognition by teachers other significant persons. In addition, sales were influenced by the extent of the artist's self-confidence. Moreover, the expectation of inferior career chances leads women avoid certain professions resulting in gender segregation. Breen and Garcia-Penalosa (2002); Filippin and Ichino (2005) showed that the beliefs about once career prospects are updated based on the success probabilities of past generations leading to gender specific career choices. As a result, female success
should be highest in areas where they found most incentives and least resistance.
Gender inequality becomes most apparent when, given equal quality of the artworks, a female artist's work is judged to be inferior than a male artist's work. In an experiment conducted by Pheterson et al. (1971) 120 college women had to evaluate the quality of a number of paintings. Half of the subjects was told that the artist was male and the other half that the artist was female. While in one scenario subjects were judging artworks of contest entrants, in the second scenario subjects were evaluating the quality of artworks of contest winners. In case of contest entrants, female paintings received a poorer judgment, while in the case of contest winner, women received equally high judgments as men. A potential explanation is that the likelihood for men to succeed is higher, which results in lower valuation for non-established female artists but overvaluation for established women who made it despite lower probability of success. The finding indicates that women encounter obstacles during their establishments as artists due to gender-related prejudicial attitudes specifically by other women. This is not the case for already recognized accomplishments pointing to a vicious circle for aspiring female artists in that unbiased valuation will only take place upon public certification. These findings are in line with the experemintal study by Goldin and Rouse (2000) where female musicians were more likely to be hired in blind auditions for symphony orchestras. Similarly, Cowen (1996) reports that the market actually did not discriminate female artists and that female artists often fetched higher prices during their lifetimes than male artists did. He argues that that it was the personal environment that discouraged women to acquire the required skill to become good artists. There is anecdotal evidence showing that established female artists such Frida Kahlo, Georgia O'Keeffe, Lee Krasner or Helen Frankthaler all enjoyed important support from influential artist fathers, teachers or husbands. It facilitated their entry into the world of art through access to relevant artistic circles. Many significant female artists indeed entered the market due to marriage or birthright (Cowen, 1996; Piirto, 1991). Several studies report that it is harder for women to find access to critical informal networks and social circles as these are often "old boys networks" (Ruth Eikhof and Warhurst, 2013; Gill, 2002; Gregory, 2009) and also benefit to lower extent from these than men do (Forret and Dougherty, 2004). These results are in line with a recent working paper by Cameron et al. (2017). Based on a sample of over 4,000 Yale School of Art graduates, the authors show that while female artist experience more difficulties entering the market they reach higher sales prices at auction conditional on accessing this market. They explain their
results with the presence of a higher bar for female artists and conclude that gender discrimination happens on institutional rather than market level.

Within the overall sphere of cultural and creative industries, empirical evidence on gender-based reward discrepancies is is also rather sparse. In their longitudinal study on the careers of male and female actors, Lincoln and Allen (2004) found that female actors receive fewer roles than male actors even when controlling for the number of Academy Award nominations. Moreover, aging seemed to have a more detrimental effect on the women's acting careers rather than the male actors' career paths. This 'double jeopardy', the adverse interaction effect of gender and age on wage, was also emprically shown by De Pater et al. (2014) in their study on wage differences among top hollywood movie stars. Bielby and Bielby (1996) found similar results for the case of a sample of screenwriters over a period of 60 years. However, these results might not necessarily apply to the world of art. While the actor's aesthetics next to their acting skills are of importance in the film industry, it is less likely to be the case for the fine art market.

The domain of professional sports is similar to the art market in the sense that it is also characterized by the superstar effect where a few top performers are able to extract most of the earnings in the industry. However, compared to the art market performance criteria are more transparent and therefore evaluation is expected to be merit-based leaving less space for discrimination. Nevertheless, remuneration for male and female athletes seems to be unequal as found in a study by Kahn (1991) for the case of top tennis players. The author showed that despite higher market values of female top players, as proxied by television ratings, prize money for men was higher than for women. In academia, a recent working paper by Hengel (2017) studies abstracts of papers submitted to top academic journals in Economics. The author finds that the abstracts of female authors are on average better written than the abstracts of male authors showing that women are held to higher editorial standards during the peer review process of academic journals. This prolongs the review process by an average of six months impacting a female researcher's career.

In summary, empirical evidence in the art market literature on gender-based performance differentials is mixed. Overall, it appears that women especially encounter barriers in entering the market. While there is some evidence that reward differential decrease once women establish themselves as artists, so far a lack of data inhibited the development of a clear understanding of gender effects in the art market and other areas characterized by superstar effect.

## 3 Data

### 3.1 Sample

The data set employed in this study was provided by Artnet AG (Artnet thereafter). The Berlinbased company is an online platform offering trading as well as research and analytic services within the art market. Their price database dates back to the year 1989 and has over ten million price quotation records. ${ }^{4}$ Artnet collects all art auction transactions which reach a hammer price equal to 500 US Dollars and above. As a result, our data set can be considered to represent the population of art auction transactions worldwide.

In this analysis, we focus on the fine art sector. The category includes photography, prints and multiples, works on paper, paintings, installations, design objects and sculptures amounting to $6,140,774$ auction transactions. In order to increase the homogeneity of our sample, we exclude installations. The market for installations is slightly different from the market for other more traditional object types as installations are more difficult to maintain, store and exhibit for collectors. Furthermore, as Artnet gradually increased the comprehension of its price database between 1989 and 2000, we restrict our sample period to the years 2000 (January) to 2017 (April) resulting in a very high degree of completion.

The database provides information on transaction characteristics including the name of the auction house and its location, the date of the sale, the lot number, the price estimate of the auction house and the hammer price in US Dollars before transaction costs. We deflate all prices using the US consumer price index (CPI) provided by the OECD using 2017 as our base year. ${ }^{5}$ With respect to the artists' attributes, the database records her name, date of birth, living status and nationality. On artwork level, we have information on the title of the work, its size and object type (as mentioned above). Additionally, we categorize all auction transactions into movements based on the birth year of the artist. Consistent with the classification in the Tefaf report, we distinguish between Old Masters and Impressionists (1250-1874), Modern (1875-1910), Post War (after 1911 and deceased) and Contemporary (all living artists). The artworks where the artist's birth year was not available are subsumed under "other". We do not consider artists born before 1250. It is important to mention that while we have artists in our sample from different artistic movements and generation, we observe

[^4]their sales only in the time period from 2000 until 2017. This implies that while opportunities for these artists differed across time, we do not expect the perceptions of buyers with regard to gender performance differences to vary too much during the period of the past 17 years.

Our variable of interest is the artist's gender. Since Artnet's price database does not indicate the gender of the artists, we identified female artist by matching them to a name list. In order to ensure accuracy and increase the homogeneity of the artists in our sample in terms of opportunities such as access to resources and education, we focus on Western artists who are based in Europe and North America (the US and Canada). Whenever there were two nationalities attributed to an artist, she was included in the sample if either of the nationalities was European or North American (e.g. the male artist Zao Wou-Ki who is French-Chinese). We combined two name lists provided by the US Social Security Administration ${ }^{6}$ and the German computer magazine Heise ${ }^{7}$. The former list contains North American baby names, while the latter provides a name dictionary with a focus on European names by country. In cases where the name was unisex (e.g. Jessy, Joan and Kim), we manually researched the identity of the artist. Instances where the artist consisted of more than one person (e.g. Christo and Jeanne-Claude) were dropped from the sample.

As a result, we are left with a sample size of $4,387,393$ observations. We drop observations where information on the dimension (size) of the object is missing which is the case for 58,166 transactions. Lastly, we exclude bought-in lots from our main analysis. ${ }^{8}$ Our final sample consists of 2,677,190 auction transactions. To the best of our knowledge, this represents the largest and most comprehensive art market auction transaction sample so far employed in a study.

Additionally, we have exclusive access to primary market data provided by Artnet. Primary market data identifies which artists are represented by which galleries and is highly confidential and therefore difficult to obtain. As a provider of art market services, Artnet provides also an online platform for art galleries to sell their work. This data set will be applied to examine the presence of entry barriers into the secondary market for female artists. It contains the name of the galleries and the names of the artists they represent as well as the artist's year of birth over the time period from 2000 until 2017. Due to disclosure limitations, we pool the data and treat it as a cross-section instead
${ }^{6}$ The list is available at https://www.ssa.gov/oact/babynames/limits.html.
${ }^{7}$ The list is available at ftp://ftp.heise.de/pub/ct/listings/0717-182.zip.
${ }^{8}$ In auctions, a buy-in takes place when an artwork is not sold as it fails to meet the seller's reserve price. The buy-in rate in our sample is $37.73 \%$ ( $1,622,019$ observations) which is in line with the commonly observed buy-in rates in auction sales.
of a panel. In total, there are 1,281 galleries in Artnet's international gallery network representing 15, 121 unique artists. Again, we only focus on Western artists. Furthermore, as we are interested in the transition from the primary to the secondary market we restrict our sample to the population of living (contemporary) artists. This leaves us with an overall sample of 4,754 artists.

The following subsection will introduce the properties of our data set and provide some first evidence for gender differentials within our sample based on univariate analysis.

### 3.2 Descriptive Statistics

Table illustrates the extent of the concentration within the secondary market based on our data for the whole sample period (2000 until 2017). It depicts for different shares of the market (in terms of value) and the percentages (numbers) of artists who account for it. First, the market is highly concentrated with only $2.2 \%$ of the artists being responsible for $90 \%$ of overall sales. Second, artwork sales of female artists amount to only $3.4 \%$ of the total auction market. Third, whereas the female segment is smaller in size it is more concentrated than the male market. While $19.9 \%$ of all male artists are responsible for $99 \%$ of the sales value, only $15.5 \%$ of all female artists occupy this share within their respective segment. These numbers suggest that the art auction market resembles a superstar market where rewards are concentrated among a few individuals. This appears to be amplified for the segment of female artists.

Table 2 shows the summary statistics for auction prices for men and women with detailed statistics by artistic movement, object type, region and living status. Overall, $96.1 \%(2,572,346)$ of all artworks sold at auction can be attributed to male artists. Hence, the proportion of female artworks in terms of volume is slightly higher ( $3.9 \%$ ) than their share in terms of value ( $3.4 \%$ ) in our sample. Figure 1 shows how the total sales value and volume developed for both genders over the sample period as well as over different generations. We chose these two dimensions since while attitudes toward gender might not have changed profoundly over the last 17 years, the market might perceive gender differently across artist generations due to the improvement of conditions for women pursuing an artistic career. As shown in Figures 1(a) and 1(c) sales volumes have clearly increased for men and women with a larger relative increase for women. While female artists increased sales volume by a multiple of 1.95 (from 3,714 artworks in 2000 to 7,247 artworks in 2016), male artists increased sales by a multiple of 1.68 (from 97,807 artworks in 2000 to 164,936 in 2016). Similarly, total sales values
have increased for both genders despite a dip following the financial crisis. From the year 2000 until 2017, female artists increased sales value by a multiple of 6.0 while male artists only increased sales by a multiple of 2.8 . Nevertheless, female artists remain a small fraction of the overall market in terms of volume and value ( $4.2 \%$ in terms of volume and $5.0 \%$ in terms of value in 2016). For both genders, sales numbers highly increased for artists born after 1875 as depicted in Figures 1(b) and 1(d). This is more pronounced for female artists and is likely to reflect a higher supply of contemporary artworks and points to lower entry barriers for female artists born in later generations.

With respect to the number of artists, men clearly dominate the auction market occupying $95.2 \%$ of the market. While there are 110,938 male artists, there are only 5,612 female artists. The proportion is highest for Contemporary art ( $9.3 \%$ are female) and smallest for the Old Masters period ( $2.9 \%$ are female). Figure 2 shows the evolution of the number of male and female artists during the sample period and over generations. From Figures 2(a) and 2(c) which depict the number of distinct male and female artists in every year, we can observe that there is an increasing trend for both groups over the years. However, the trend is stronger for the female sub-group with an almost three-fold increase from 165 artists in 2000 to 456 artists in 2016. The number of male artists at auction per year less than doubled from 4,303 to 7,815 artists over the same time period. As a result, the male-to-female ratio improved by $40 \%$ over time from 0.03 in 2000 to 0.05 in 2017 (see Figure A2 in the Appendix). This trend is also reflected in Figure 2(d) which shows a steady increase in the number of female artists over the generations with a clear peak for the generation that was born between 1975 and 2000. The number of male artists remains rather stable for the generations born after the year 1875. The rising market entry by female artists points to a potential improvement in conditions and higher market acceptance making the artist profession more attractive for women.

An interesting observation is that while the average prices of female artworks are below the average price for male artworks ( $\$ 39,065$ versus $\$ 45,614$ ), the median price is with $\$ 3,931$ higher for women than for men ( $\$ 3,649$ ). This is also reflected in Figure 3 which shows how these numbers have evolved over time and through generations of artists. In Figure 3(a) we can observe that mean artwork prices are consistently higher for men, whereas median prices (3(c)) are higher for women after 2002 with a widening gap after 2011. The hedonic price indices based on the respective time (year) dummies for both genders in Figure A1 in the Appendix show that sales of female artists have overall outperformed male artists (1(a)). However, this seems to be driven by artists from older
generations since the financial performance of contemporary female artist (1(b)) appears to be worse than the performance of contemporary male artists.

Paintings are most frequent object type in our data set for both genders while Photographs are the least frequent object type. Mean artwork prices are lower for women for Paintings and Works on Paper while median prices are only lower for Prints and Multiples.

In terms of national residency, it is noteworthy that mean artwork prices of female artists are slightly higher in North America ( $\$ 58,929$ versus $\$ 58.234$ ) and a significantly higher in Eastern Europe ( $\$ 68.258$ versus $\$ 40.758$ ). Only in Western Europe median prices for female artists are lower than for male artists.

With respect to living status, the share of artworks by deceased artists is lower for the female sub-sample ( $64.9 \%$ ) than for the male sub-sample ( $78.5 \%$ ). Furthermore, artworks by both living and deceased female artists fetch higher mean and median prices than artworks by male living and deceased artists.

Lastly, Tables A1 and A2 in the Appendix provide an overview of the top 25 male and female artists and reveal some first insights on the rank of female artists in the market. With a sales value of $\$ 392,962,816$ the highest selling female artists, Joan Mitchell, does not reach the sales value of any of the male artists in the top 25.

In summary, the univariate analysis reveals three important facts about gender differences in the secondary art market. First, with a share of less than $4 \%$ female artists are extremely underrepresented but relatively more concentrated in terms of sales value in the secondary art market. Second, median prices appear to be higher for women. This might be indicative of a selection mechanism where a higher bar is applied to female artists admitting only the most talented ones. Third, it appears that those women, who do break through the initial barrier to the market, still lag behind top male artists in terms of sales value. In the following section, we will perform an in-depth multivariate analysis in order to investigate the presence of structural barriers for female artists in the secondary art market.

## 4 Empirical Analysis

### 4.1 Selection into the Art Market

Our first hypothesis is that female artists are less likely to enter the secondary art market than men. As shown in the descriptive statistics, female artists are highly underrepresented in the secondary market with a share of less than $4 \%$. At the same time, it is reported that the number of female students pursuing MFA degrees is not below the number of men. This indicates that there appears to be a large drop out rate of women between these two career stages. However, it might be the case that not all students attending fine art schools are interested in pursuing professional careers as artists. Therefore, in order to make conclusions with respect to the mobility of female artists, we need to observe the share of women present in the primary market where less established and younger artists are represented by galleries.

If an artist is present in the primary data set, it means that she is represented by at least one gallery during the sample period. Having gallery representation is the first crucial step in an artist's career after completing her education. A gallery provides the artist with access to its network of buyers as well as marketing activities to improve her visibility in the art market. The gallery is compensated through a sales commission on the price of the artwork. While galleries can represent emerging as well as more established artists, good representation is particularly important for new, unknown artists. Auctions are considered the secondary market where sellers mainly consist of collectors who previously acquired an artwork in the primary market (or in the secondary market). Reasons for selling an artwork at auction can be liquidity related (the famous disease, death and divorce) or the fact that an artwork has increased in market value. In the latter case the owner wishes to realize this positive financial return. Collectors are more likely to offer an artwork for sale at auction if it has appreciated in value and if there is demand for it by other buyers. Hence, if an artist is not traded at auction it suggests that her artworks have not gained sufficiently in value to make it attractive for their owners to sell them.

In order to determine how many male and female artists move from the gallery to the auction market, we check whether the artists in the contemporary primary market sample are also present in our main (auction) sample of living artists. Table 3 shows that out of 4,180 male artists, $96.9 \%$ (4,050 artists) can also be found in the secondary market. However, only $93.0 \%$ ( 534 artists) out of the 574
female artists made this transition. The difference in proportions test is statistically significant on a $1 \%$ significance level. It is also notable that the share of women decreases from $13.7 \%$ in the primary market to $11.6 \%$ in the secondary market within this sample. This amounts to a drop of $15 \%$. The result of this univariate analysis provides us with a first evidence for a barrier that precludes female artists from entering into the secondary market.

The fact that female artists are not as likely to reach the secondary market as male artists suggests that there is a selection mechanism in place which admits only female artists with certain attributes to the market. These might be the most talented and pertistent women in the sample. If this is the case, we should observe in our analysis that being female is positively correlated with the value of total sales. Descriptive statistics in Table 3 show that the average total sales value for men is twice as high as the mean total sales for women ( $\$ 3.4$ million versus $\$ 1.5$ million). The total sales value is defined as the sum of the value of all sold lots at auction over the sample period (2000 to 2017) per artists in 2017 US Dollars. In order to analyze the performance of women in a multivariate setting and to control for a potential selection bias, we employ a Heckman sample selection correction as suggested in Heckman (1977) on the entire primary market sample with the log of the total sales as our dependent variable. The model takes the following form:

$$
\begin{align*}
Z_{j} & =\alpha_{1}+\delta_{1} D_{j}+\lambda_{1} A_{j}+\epsilon_{1 j},  \tag{1}\\
\log S_{j} & =\alpha_{2}+\delta_{2} D_{j}+\lambda_{2} A_{j}+\epsilon_{2 j}, \quad j=1, \ldots, N_{a} \tag{2}
\end{align*}
$$

where equation (1) is the selection equation and equation (2) is the outcome equation. $Z_{j}$ is a binary variable that takes the value 1 if an artist $j$ participates at auction and 0 otherwise. $\log S_{j}$ is the $\log$ of the total deflated sales value for each artist $j$ from the population of $N_{a}=4754$ living artists in our sample. $D_{j}$ denotes the discrimination coefficient which is a gender dummy taking the a value 1 whenever the respective artists is a woman. $A_{j}$ is a $1 \times 92$ vector that denotes the artist characteristics including the artist's nationality ${ }^{9}$, his or her year of birth as well as a dummy for every gallery an artist is represented by as a gallery's reputation is known to have a high impact on

[^5]an artist's success. ${ }^{10} \delta_{1}, \delta_{2}, \lambda_{1}, \lambda_{2}$ and $\kappa$ are parameters of the model. $\alpha_{1}$ and $\alpha_{2}$ are constant terms. $\log S_{j}$ is only observed if $Z_{j}$ equals 1 . As we are dealing with a cross-sectional data set we do not include any time-fixed effects. Following Heckman procedure, we assume that $\epsilon_{1 j}$ and $\epsilon_{2 j}$ are error terms which follow a bivariate normal distribution from which can be derived a scale parameter $\sigma$ and a correlation coefficient $\rho$. While we do not make use of any exclusion restrictions (all variables in the selection model (equation (1)) are also included in the outcome model (equation (2)), we rely on the non-linearity of the data to identify the model. The correction term will not be perfectly correlated with the covariates since the Mills ratio is estimated using a non-linear probit model.

The results of the selection and outcome model are shown in Table 4. The first important observation is derived from the results of the selection model. It provides evidence for a small but statistically significant barrier for female artists at the transition from the primary into the secondary market. The presented coefficients in the selection model are the marginal effects at the mean. The coefficient on the female dummy indicates that given their characteristics, female artists are $2.2 \%$ less likely to participate at auction compared to men. For example, a female artist might have interrupted her artistic career which would negatively impact the market demand for her existent artworks. This can also be the result of self-selection whereby female artists decide to cease their artistic endeavors in anticipation of less success. It could also be the case that galleries underinvest in female artists as they estimate their likelihood to succeed as lower. An alternative explanation might that buyers of female artworks differ from other other buyers with respect to their buying motive. These buyers could prefer to hold on to their purchases being less interested in realizing financial returns. While we cannot exclude this alternative explanation, there is currently little theoretical ground to support such a conjecture.

Furthermore, the correlation coefficient $\rho$ is relatively large and highly statistically significant. This implies that there is indeed a correlation between the error terms of the selection and the outcome model. Failing to correct for the selection mechanism in the outcome model would result in an upward bias in the coefficient on the female dummy. In other words, female artists who transition into the auction market are different than the average female artists in the sample. Nevertheless, after correcting for the selection bias in the Heckman model, the female dummy is negative and

[^6]statistically significant.
Overall, the magnitude of the chance with which female artists are less likely to transit from the primary into the secondary market ( $-2.2 \%$ ) appears to be low and provides evidence of increased equality between genders in recent years. However, it is to keep in mind that this estimate is on the lower bound given that we are only considering contemporary artists who are represented by US galleries. It is also worth mentioning that Artnet's gallery network does not capture the whole population of galleries. It is likely that our gallery sample is rather biased towards more successful artists as being a member of an online gallery network requires resources smaller galleries might not possess. As a result, it might be the case that the artists in our gallery sample have a larger chance to progress into the secondary market. This implies that our findings are on the conservative side and that the actual share of female artists not transitioning into the secondary market is even likely to be higher. Another concern is that the relationship between gallery representation and the auction participation works in the opposite direction and that good auction results lead to a higher probability of being represented by a gallery. Nevertheless, this would still not explain why women have a lower chance of being represented by a gallery given their auction performance.

### 4.2 Performance at Auction

In the previous section we have found that based on our gallery sample female artists are less likely to progress from the primary into the secondary market. We will now turn to our main (auction) sample to investigate the overall performance of female artists on artwork level at auction. The basic regression model has the following specification,

$$
\begin{equation*}
\log P_{i t}=\alpha+\psi W_{i}+\beta X_{i}+\eta H_{i}+\tau_{t+} \epsilon_{i t}, \quad i=1, \ldots, N ; \quad t=1, \ldots, T ; \tag{3}
\end{equation*}
$$

where $\log P_{i t}$ indicates the $\log$ of the real price of an artwork $i$, which is sold at a given time $t$. $N=2,677,190$ artworks in our sample over $T=72$ seasons between 2000 and 2017. $W_{i}$ denotes the discrimination coefficient which is a gender dummy taking a value of 1 whenever the respective artists of a given artwork, $i$, is a woman. This regression specification estimates the differences between the actual sales price for an artwork of a female artist and the value of an artwork by a male artist with the same characteristics. All artwork characteristics are captured in $X_{i}$, a $1 \times 276$ vector that includes the object type (the base category are paintings), the auction house where it
was sold and the size of the artwork. ${ }^{11} H_{i}$ is a $1 \times 5$ vector that denotes the artist characteristics of a given artwork $i$ including region of the artist's nationality (the base category is North America) ${ }^{12}$ and a dummy for the living status of the artist at the time of the transaction (the base category is deceased). Due to collinearity between the artist names and the gender dummy, we exclude artist fixed effects from the regression. $\tau$ represents time fixed-effects for the years 2000 until 2017. $\psi, \beta$ and $\eta$ are time-independent parameters. $\alpha$ is a constant term. Lastly, $\epsilon_{i t}$ denotes the error term.

Table 5 reports the regression results when estimating parametrs using the OLS methodology. While the first two specification involve the full sample using the real and the nominal price as the dependent variable, we also run seperate regression for all four movement sub-samples with the real price as the dependent variable. The highly statistically significant female dummy coefficient in our base regressions shows that artworks by female artists are on average $4.4 \%$ more expensive than the artworks of male artists given the quality of the artworks. While this difference appears to be rather small, this depicts merely the average effect. It is also consistent with findings by Bertrand and Hallock (2001) who studied gender salary differentials for the case of top executives. This result implies that there is a premium on artworks created by women which is supportive evidence for the presence of a selection mechanism whereby female artists that make it to the secondary art market are on average better than male artists. It could also be indicative of a potential supply squeeze. Due to the limited supply of high-quality female artworks, collectors are willing to pay a premium for these lots.

All other coefficients are in line with expectations. Sculptures are the most expensive objects, while prints and multiples display the highest discount relative to paintings. Artworks of artists from Southern Europe sell highest. This is not surprising given that many of the top artists such Picasso, Modigliani, Miro and Fontana originate from there. Lastly, there is a premium on deceased artists. The R-squared of the regression is 0.42 which is within the usual range for hedonic models in the field of art market economics (Ashenfelter and Graddy, 2002). All coefficients remain unchanged independent of whether the nominal or the real artwork price is used as the dependent variable.

Male and female artists in different time periods were subject to different conditions especially with respect to access to education and the general acceptance of women as creators of cultural goods

[^7]and part of the workforce. Assuming societal barriers as the only source of performance differences between men and women, our base regression results might pick up unobserved quality difference between the artworks produced due to unequal opportunities granted to women. For instance, less support for female artist in the Old Masters movement could have led to an even higher bar for women during this time period reinforcing a potential selection mechanism. In turn, this would imply that women who succeeded to pursue careers as professional artists had to be on average better than male artists resulting in a higher demand for these lots. As opportunities and beliefs held in society with respect to gender roles have shifted throughout time, the selection mechanism is expected to become less pronounced resulting in a convergence in the supplied quality between artworks produced by men and women in later time periods. As a result, any observed performance differences at auction between the genders are expected to be to a lesser extent due to factors related to differences in access to opportunities.

In order to test this proposition, we estimate the model specified in equation (3) for each artistic movement separately. The last four specifications in Table 5 present the results. Interestingly, the coefficient on the female dummy is positive and statistically highly significant for each movement with the exception of Contemporary art where we observe a negative and statistically significant coefficient. The Post War Era yields the largest premium (14.9\%) for female artworks. While it is difficult to imagine that opportunities were worse for female artists in the mid-20th century than in the mid-19th century, it could be the case that this era produced a small number of female artist considered large superstars (e.g. Agnes Martin, Helen Frankenthaler and Joan Mitchell) for which competition among buyers is high. The discount on contemporary female lots is more likely to be driven by a gender bias present in the market given the improved opportunities for women pursuing an artistic career manifested in a relatively higher proportion of women in this period (9.3\%). However, it is also indicative of a lower bar and a larger demand for female art whereby also artworks of lesser quality enter the secondary market. ${ }^{13}$

Even though our data set contains artworks created by artists in different time periods, all sales of these works take place over a time period of about 18 years (2000 to 2017). While we do not expect large shifts in the market attitude towards female artists or strong differences in the quality

[^8]of the artworks by men and women available in the market, we are interested in investigating the persistence of the difference in performance between the genders found in our baseline regression in Table 5 across time. We split our data into four different time periods for which we run separate regressions. The results are shown in Table 6. For all four periods a premium for female lots persists ranging from $1.9 \%$ to $7.4 \%$. The premium appears to be smaller for the years after 2010. However, this might be due an increased supply of artworks in the market in later years by female artists (see Figures 2(a) and 2(c)). Since we observed a discount on female Contemporary lots (Table 6), Table 7 separately reports the results for the sub-sample of Contemporary artworks for the four time periods. Interestingly, it appears that the discount on female lots intensifies throughout time. While the marginally statistically significant discount amounts to $3.5 \%$ in the period from 2000 until 2004, it increases to $12.6 \%$ for the years from 2015 until 2017. These result also reject the participation rate hypothesis which states that extreme performance outcomes are less likely for women for statistical reasons as they are fewer in number. It aims to explain why women are excluded from top ranks in occupations with a high concentration of men. However, as our results show, increasing the ratio of women or lowering their barriers to entry does not defeat a gender gap in performance. This suggests that these additional female artists fall into the lower price quantiles lowering the average performance and could be interpreted as a sign of lower barriers for female artists.

Another way to better control for the unobserved quality characteristics of the artworks which are not explicitly captured by our hedonic variables, would be to modify the dependent variable (the artwork price) by dividing it by the mid-point of the auction house pre-sale estimate. ${ }^{14}$ This way, we would analyze whether gender can explain the auction house's estimation error which is equal to the deviation of the final hammer price from the auction house price estimate. However, this would assume that auction house estimates are unbiased measures of quality. We believe that this is unlikely given that auction house experts incorporate buyer preferences and tastes in their valuations of the artworks. In addition, there are over 1,500 auction houses in our sample with diverging valuation procedures. Lastly, in this study we are interested whether a gender bias is present in the overall art market and not only on institutional level. For robustness, we performed a regression using the model specification in equation (3) with the log of the price scaled by the auction house pre-sale estimate as the dependent variable. The results can be found in Table A4 in the Appendix. The female dummy

[^9]coefficient together with most of the other coefficients on our hedonic variables becomes statistically insignificant. This means that auction houses are on average able to provide a reliable estimate of the value of an artwork and account for a potential gender bias among buyers.

Overall, the analysis in this section yields two important findings. First, it appears that women are less likely to transit from the primary into the secondary market. While small in magnitude, this impaired mobility is suggestive of a selection mechanism that sets a higher bar for women and admits only the most persistent and/or most talented female artists to the auction market. Second, there is a premium on artworks for those women who break through this initial hurdle. This might be the consequence of the higher bar and a limited supply of high-quality female lots. This effect is in particular driven by artworks of older generations where different possibilities for men and women prevailed. At the same time, we observe an over time increasing discount for Contemporary female lots which suggests lower barrier for women in recent times. Given the low amount of female artists and the superstar effect in the art market, our conjecture is that the premium observed for older generations of artists is driven by a small number of female artists whose artworks demand very large prices. These top artists could be causing a supply squeeze as in a "winner-take-all" market demand will be concentrated around these few individuals. Furhet, a potentially skewed distribution of sales is not taken into account by OLS estimation which focuses on the average effect. Therefore, the next section will take a closer look at the distribution of sales for male and female artists.

### 4.3 Distribution of Rewards

The analysis above showed that the share of female artists decreased as they moved from the primary into the secondary market. This entry barrier appears to have caused a selection bias in the population of female artists traded at auction. The result is that it looks like women outperform men in terms of prices fetched at auction. However, we hypothesize that this average effect results from a supply squeeze for the most popular female artists who attract the highest demand. These women constitute the superstars in the "winner-take-all" market and drive the observed average premium. Results on artwork level are distorted if a large amount of artworks is sold by a small amount of female artists located on the top of the price distribution.

Table 8 depicts the distribution of lots of male and female artists. Overall, we can see that a smaller amount of female artists accounts for a relatively larger amount of lots in terms of volume
and value than it is case for male lots. For instance, $95 \%$ of the female lots sold at auction stem from $36.7 \%$ of female artists accounting for $99.5 \%$ of the overall value of female lots. For the male sample, $95 \%$ of all lots are covered by $40.0 \%$ of the male artists who absorb $99.0 \%$ in terms of artwork value. Given the difference in the concentration between the male and the female market, the artwork level OLS is likely to be not informative about the true performance of female artists at auction.

In order to correct for the large number of lots by the most expensive artists, we estimate equation (3) using weighted-least-squares (WLS). The applied weights equal the inverse of the square root of the total number of artworks sold per artist at auction throughout the sample period. This transformation results in an equal weight for every artists in our sample. The results of the WLS regression are presented in Table 9. As expected, the female dummy coefficient turns negative and now yields an average discount of $10 \%$ on female lots given the characteristics of the artworks. Again, this result remains unchanged when the nominal price is employed as the dependent variable. This finding lends supportive evidence for the conjecture that the positive coefficients on artwork level derived in the previous section is due to the presence of a small number of very popular female artists who attract the largest demand and the highest prices from collectors in a "winner-take-all" market.

As in the previous section, we perform a robustness check for the different sub-samples consisting of the four artistic movements. The discount on female lots observed in base specification seems to be driven by Post War and Contemporary lots which yield a female dummy coefficient of $-13.1 \%$ and $4.2 \%$ respectively. While the price discount becomes smaller compared to the OLS baseline regression in Table 5 for Contemporary female artists, it flips the sign for the case of women attributed to the Post War era. It appears that particularly in this movement buyers compete for a very small number of female superstars (such as Joan Mitchell and Agnes Martin) with a large discount for the average female artist active during this period. The Post War era is characterized by abstract geometric forms which might more likely be associated with male attributes. The relative reduction in the magnitude of the discount for Contemporary female lots implies that this movement in particular dominated by a number of women with a large amount of lower priced artworks which are less desired by buyers. However, even if we take this distribution into account, the discount persists. The female dummy coefficient for Modern lots is negative but statistically indistinguishable from zero. Female Old Masters lot retain a premium, however the coefficient is only statistically significant on a $10 \%$ level. Thus, it appears that the premium found in the previous section is likely to be driven by a
supply squeeze for a few superstar female artists in the case of Post War art while it was due to a higher quality standard in the case of Old Master works.

Given that women are less likely to gain access to the secondary art market resulting in an upward bias and the fact that the art market is a "winner-take-all" market, we further investigate the concentration within the female segment of the market. Table 10 shows the percentages of male and female artists at every quantile of the sales value distribution on artist level. As defined in the section above, the sales value equals the sum of the value of all sold lots throughout the sample period per artist. The most interesting observation is that the female sub-market is more concentrated at the top (top 10\%) and less concentrated at the bottom (bottom $50 \%$ ) than the male sub-market. The latter effect becomes more amplified the further we move down the sales value distribution. While an expected share of $10.1 \%(5,650)$ of the male artists can be found in the top $10 \%$ of the sales value distribution, only $7.5 \%$ (178) of the female artists are located there. At the 50th quantile of the sales value distribution, only a total of $38.9 \%$ of the female artists can be found as opposed to an expected share of $50.6 \%$. Moving further down the sales value distribution, $9.7 \%$ of all male and as many as $15.9 \%$ of the female artists are situated at the bottom $10 \%$ of the sales value distribution. Overall, this implies that female artists are more likely be found at the bottom of the sales distribution. The superstar effect wherein a small number of individuals absorbs all industry rewards (Rosen, 1981) applies even more to the female sub-group than to the male segment.

### 4.4 The Superstar Effect

If being traded in the secondary market is not sufficient to signal quality and legitimacy for female artists, we expect that should materialize in a glass ceiling on the top of the market where the largest rewards are concentrated.

Table 10 does not only exemplify the concentration of sales within the female sub-segment, but also provides first critical evidence for a second barrier for female artists in the market. In the 99.97th quantile of the sales value distribution no single female artist can found. This quantile corresponds to a market share of $40 \%$ in terms of value which entirely accrues to a core of 40 top male artists. As the most expensive female artist, Joan Mitchell, can be found in the 99.96th quantile. With $\$ 393$ million in total sales, she is ranked 43rd in the list of top artists.

Furthermore, Table 11 shows for different quantiles of the sales value distribution the respective
brackets for male and female artists as well as the number of artists per bracket. The key takeaway is that the sales value is significantly lower for female artists than for male artists in every quantile with exception of 99th quantile where the sales value bracket is $\$ 9,342,266$ for male artists and $\$ 12,382,016$ for female artists. However, at the very top, namely at the 99.91th quantile, the sales value per artist for men elevates again above the sales value level of women. Moving from $\$ 9.3$ million in the 99th quantile to 176.8 \$million in the 99.1 th quantile, is a sizable jump. At this sales level 99 male artists and a mere of 5 female artists can be encountered. While the overall 1:20 male-to-female ratio is preserved at this quantile, the increase in sales values is disproportional. It represents the part of the distribution where the superstars of the art market are located who absorb the largest chunk of the rewards. This univariate artist level analysis shows that in order to reach and overshoot the sales level of male artists, a woman needs to be at the top of the distribution. At the same time, she is precluded from entering the league of the superstars of the art market which appears to be reserved for the male population of artists.

In the following step, we aim to investigate whether taking into account the skewed distribution of prices in the art market will open up a more granular view on gender differences in the art market. Capitalizing on our comprehensive data set that allows us to dig into the tails of the price distribution, we estimate parameters of equation (3) with a quantile regression technique as laid out by Koenker and Bassett Jr (1978). Quantile regression models consider every price segment separately focusing on parts of the distribution other than the conditional expectation. Table 12 presents the regression results and offers a very clear perspective on gender effects in our sample and are in line with the findings from the univariate analysis. For illustration, the female dummy coefficient is plotted in Figure 4. In the lowest quantile, there is no significant differences in terms of prices between male and female artists. After the 25th quantile, we observe a premium on female artworks which steadily increases from $4.5 \%$ to $6.7 \%$ in the 95 th quantile. A discount of $1.8 \%$ emerges for artworks by female artists at the 99th quantile and amplifies to $9.1 \%$ at the 99.9 th quantile, which represents the very top of the secondary art market. This is supportive evidence for the presence of a glass ceiling that precludes women from participating in the high-end of the art market.

Furthermore, even though ? argues that topics are rather similar within artistic periods as opposed to within gender, other potential explanatory variables that could be correlated with the female dummy are certain colors or themes. For instance, it could be the case that female artists are
more likely to focus on family themes in their artworks. These topics might in turn be valued higher or lower by the market. As these variables should be artist specific, we make use of a two-stage regression. We perform this analysis in two steps due to the correlation between the artist and the gender dummies. In the first step, we include artist fixed-effects in our regression on the logarithm of the real price in addition to the other hedonic variables specified in equation (3). We extract the residuals from this regression and run a second regression where the residual is the dependent variable and the gender dummy is the independent variable. The results are reported in the Appendix in Table A5 and show that even after considering the artist as an explanatory variable, the gender dummy is still highly statistically significant. With exception of the coefficient in the 99.9th quantile which is indistinguishable from zero, the pattern of the coefficients along the quantiles is consistent with the findings in Table 12. However, as previously mentioned, the gender dummy and the artist are correlated variables. Therefore, the results of this two-step residual regression are to be viewed with caution.

Overall, the preceding analysis showed that at the very top of the market the idea of the male artistic genius still prevails. The "winner-take-all" effect in the art market appears to create a structure which precludes female artists from reaching the top of the market.

## 5 Concluding Remarks

This is the first large-scale study to provide strong empirical evidence for how a market structure characterized by the superstar effect can impact the career mobility of an underrepresented group. Using a sample of data that is only fractionally smaller than the true population enables us to accurately describe and analyze gender performance differences in the secondary market for fine art. The analysis goes beyond establishing the average effect of being a woman on the price of artworks. Instead, we closely look into the upper tail of the sales value and price distributions where most of the rewards are concentrated.

First, we show that female artists are still highly underrepresented in the primary (gallery) as well as in the secondary (auction) market. While the share of women in art schools acquiring MFA degrees is reported to be equal to the share of men, we encounter a proportion of only $13.7 \%$ in our primary art market sample of contemporary artists which decreases to $11.6 \%$ in the secondary market. Overall, across movements and generations, female artists make up only a share of less
than $4 \%$ in terms of number of artists as well as number of lots. This appears to be the result of higher entry barriers for women than men when transitioning from the primary to the secondary market. Specifically, results from the Heckman selection model show that women are $2.2 \%$ less likely to progress into the secondary market. This provides supporting evidence of a higher quality standard applied to female artists. In line with the higher bar explnation, we observe an average price permium of $4.4 \%$ on artwork level for female artists which is driven by artists of older generations where opportunties were less equal presumably allowing only the best and most presistent women to pursue an artistic career. These findings are consistent with the most recent working paper by Cameron et al. (2017) who find a premium for female artworks traded at auction within a sample of Yale graduates. A higher bar for women was also found in recent studies within the area of academia where journal papers by women had to be better written (Hengel, 2017) and science where patent application by women were found to be subject to higher scrutiny (Jensen et al., 2018). At the same time, our results show that the share of contemporary female artist has increased and that they are subject to a price discount. This might be indicative of lower entry barriers for women in recent years.

Second, we provide empirical evidence that the superstar effect which is a characteristic of the art market is more prevalent within the group of female artists than within the male segment. The observed average price premium for female artworks turns into a $10 \%$ price discount after correcting for the number of lots per artist. This discount is driven by Post War and Contemporary artists and implies that the observed price premium is due to a small number of female artists who account for a large share of expensive lots. Additionally, we find that the top end of the market is more concentrated in female sub-sample than in the male sub-sample. The women located at the top of the sales distribution appear to be responsible for the price premium and potentially cause a supply squeeze for their limited amount of lots. Furthermore, relatively more female than male artists are located in the lower tail of the value distribution. In every quantile of the distribution, the total sales value for men is higher than the one for women with exception of the 99th quantile. This implies that unless a female artist reaches the top, her sales will remain below the sales level of a male artist in the secondary art market. This has also been shown for the case of women in top executive positions and for women in higher salary quantiles in general (Bertrand and Hallock, 2001; Garcia et al., 2001; Kuhn, 1987).

Third, we reveal that the top end of the art market is still dominated by a core of male artists. In terms of total sales values, the 99.97th quantile which corresponds to $40 \%$ of the market by value is entirely occupied by male artists where no single women can be found. This is supported by the quantile regression results which show that within the 99.9th quantile of the price distribution a discount of $9 \%$ for female artworks emerges. This result is in line with empirical findings in other industries where the superstar effect prevails including the market for top athletes (Kahn, 1991), movie stars (Bielby and Bielby, 1996; Lincoln and Allen, 2004), high-level executives (Bertrand and Hallock, 2001) and researchers (Barbezat and Hughes, 2005; Probert, 2005) and might be the result of the given market structure which prevents the mobility of a historically underrepresented group. It appears that being traded on the secondary is not sufficient for a female artist to signal quality and establish legitimization.

Overall, it appears that gender still plays an important role in the art market. While the art industry has grown substantially over time with rising rewards for artists located at the top of the market as documented by the recent Art Basel and UBS Art Market Report (McAndrew, 2018), these rewards are still concentrated among a core that purely consist of male artists. While intitially women are subject to higher quality standards at the very high end of the market, prices are lower for an artwork conditional on the artist being female as compared to an artists being male. This can lead to a self-selection mechanism for female artists. As emprically shown by Breen and GarciaPenalosa (2002), it appears that the anticipation of lower sales might discourage women from pursuing professional artistic careers leading them to drop out of the market. Only female artists with the lower opportunity costs and highest talent might be willing to remain in the market. While this might be efficient, there is no explanation why the same filter should not be applied to the population of male artists. On the other hand, we show that the number of artworks by women traded in the secondary market has increased. The price discount on female contemporary lots can be interpreted as a sign that the bar for female artist to enter the market became lower over the years and demand for their lots has increased.

Our study provides important lessons for gender differentials in labor market outcomes within markets characterized by the superstar effect. Our results suggest that an inertia towards existing market structures can hamper the mobility of historically underrepresented groups. While in other occupational areas (e.g. orchestra auditions) blinding the identity of the individuals has mitigated
gender inequality this is less feasible in the art market. Part of an artwork's value is constituted by its historical value which is difficult to detach from an artist's identity. Therefore, it is important that institutions in the art world deviate from the current paradigm in order to establish gender equality. At the same time such non-conformance needs to be made worthwhile for institutions giving also an important role to policymakers.

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## Figures



The year 2017 is omitted in Figures a) and c) as we only use the first four months of this year. Overall, there were 35,860 artworks by male and 1,787 artworks by female artists in this year. The value of these artworks is $\$ 1,521,769,000$ and $\$ 53,611,000$ respectively. Due to missing data on the year of birth not all artists could not be allocated to a generation. Figures b) and d) omit these artists. Overall, there are 89,888 artworks by male and 2,199 artworks by female artists in this omitted category. The value of these artworks is $\$ 761,310,000$ and $\$ 7,780,000$ respectively.

Figure 1: Evolution of sales by male and female artists


The year 2017 is omitted in Figures a) and c) as we only use the first four months of this year. Overall, there were 6,171 male and 167 female artists in 2017. Due to missing data on the year of birth not all artists could not be allocated to a generation. Figures b) and d) omit these artists. Overall, 21,748 male and 1,113 female artists could not be allocated to a generation.

Figure 2: Evolution of number of male and female artists


The year 2017 is omitted in Figures a) and c) as we only use the first four months of this year. Overall, the mean (median) value is $\$ 42,436(\$ 3,681)$ for artworks by male and $\$ 30,001(\$ 4,306)$ for artworks by female artists in this year. Due to missing data on the year of birth not all artists could not be allocated to a generation. Figures b) and d) omit these artists. Overall, the mean (median) value is $\$ 8,968(\$ 1,992)$ for artworks by male and $\$ 3,542(\$ 1,182)$ for artworks by female artists in this omitted category.

Figure 3: Evolution of mean and median artwork prices for men and women


Figure 4: Coefficient on female dummy in quantile regression

## Tables

Table 1: Concentration in the auction market (2000-2017)

|  | Share of market value |  |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: |
| Share (number) of artists | $50 \%$ | $75 \%$ |  | $90 \%$ | $99 \%$ |
|  | Total value |  |  |  |  |
| All artists | $0.07 \%(80)$ | $0.43 \%(497)$ | $2.18 \%(2,563)$ | $19.67 \%(22,926)$ | $\$ 121.4 \mathrm{bn}$ |
| Male artists | $0.07 \%(73)$ | $0.41 \%$ | $(453)$ | $2.16 \%(2,401)$ | $19.89 \%(22,065)$ |
| Female artists | $0.27 \%(15)$ | $0.89 \%(50)$ | $2.41 \%(135)$ | $15.54 \%(872)$ | $\$ 4.3 \mathrm{bn}$ |

Table 2: Summary statistics for men and women

| Price | Men |  |  |  |  | Women |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | N | N artist | mean | median | sd | N | N artists | mean | median | sd |
| Overall | 2,572,346 | 110,938 | 45,614.1 | 3,647.8 | 686,070.3 | 104,844 | 5,612 | 39,065.3 | 3,930.9 | 330,635.0 |
| Buy-in rate* |  |  | 0.378 | 0.376 | 0.027 |  |  | 0.361 | 0.361 | 0.034 |
| Movement |  |  |  |  |  |  |  |  |  |  |
| Contemporary | 388,070 | 19,917 | 38,024.9 | 3,145.6 | 432,128.5 | 30,434 | 2,031 | 28,501.7 | 4,011.2 | 170,396.0 |
| Postwar | 532,238 | 34,173 | 41,046.7 | 3,090.3 | 635,913.1 | 24,280 | 1,863 | 54,261.5 | 4,120.7 | 405,304.9 |
| Modern | 819,923 | 21,281 | 51,357.5 | 3,541.6 | 860,069.0 | 34,920 | 961 | 41,331.8 | 3,700.8 | 409,739.3 |
| Old Masters | 525,405 | 19,806 | 48,503.0 | 3,776.2 | 678,066.8 | 13,781 | 595 | 29,921.3 | 3,927.8 | 223,868.1 |
| \& Impressionists |  |  |  |  |  |  |  |  |  |  |
| Other | 306,710 | 15,761 | 38,641.4 | 5,743.2 | 494,310.3 | 1,429 | 162 | 34,315.6 | 5,418.4 | 135,044.4 |
| Object type** |  |  |  |  |  |  |  |  |  |  |
| Design | 212,709 | 9,250 | 12,847.8 | 3,269.2 | 67,872.9 | 11,141 | 521 | 19,934.2 | 4,008.9 | 17,703.1 |
| Sculptures | 169,704 | 15,306 | 70,600.0 | 5,031.8 | 852,475.8 | 8,132 | 807 | 88,341.0 | 11,419.3 | 483,139.8 |
| Paintings | 1,132,403 | 78,184 | 75,343.8 | 4,903.5 | 951,020.9 | 33,064 | 3,663 | 72,025.3 | 5,141.5 | 486,107.5 |
| Works on paper | 453,729 | 36,161 | 24,542.8 | 3,090.4 | 315,270.5 | 16,477 | 1,646 | 18,470.0 | 3,796.9 | 93,930.6 |
| Prints and multiples | 477,203 | 15,050 | 2,116.7 | 2,240.6 | 172,222.1 | 19,371 | 711 | 6,629.6 | 1,895.0 | 100,688.1 |
| Photographs | 126,598 | 6,822 | 15,476.9 | 3,572.0 | 76,292.3 | 16,659 | 603 | 20,474.7 | 5,124.6 | 107,242.9 |
| Region |  |  |  |  |  |  |  |  |  |  |
| North America | 545,239 | 24,641 | 58,233.6 | 3,945.8 | 803,389.2 | 34,751 | 1,727 | 58,929.0 | 4,524.6 | 467,225.4 |
| Northern Europe | 463,192 | 19,162 | 29,560.2 | 3,032.8 | 593,015.9 | 25,195 | 1,310 | 27,827.0 | 3,624.7 | 263,459.0 |
| Western Europe | 1,099,021 | 44,143 | 43,113.7 | 3,570.9 | 594,710.6 | 35,243 | 1,673 | 24,473.0 | 3,393.6 | 140,217.4 |
| Southern Europe | 337,164 | 14,049 | 57,251.4 | 4,245.5 | 912,560.6 | 4,040 | 329 | 25,011.5 | 5,695.4 | 111,384.2 |
| Eastern Europe | 127,730 | 8,943 | 40,758.2 | 4,246.9 | 453,220.3 | 5,615 | 573 | 68,258.4 | 4,257.7 | 491,060.8 |
| Living status at time of sale |  |  |  |  |  |  |  |  |  |  |
| Deceased | 2,018,743 | 65,760 | 49,159.4 | 3,893.2 | 748,360.2 | 68,033 | 2,263 | 44,658.6 | 3,940.7 | 390,599.5 |
| Alive** | 553,603 | 47,175 | 32,686.1 | 2,863.5 | 380,339.7 | 36,811 | 3,454 | 28,728.0 | 3,909.0 | 170,957.0 |

*The buy-in rate is the share of lots of all lots offered per artist that is not sold at auction.
In total, 156,761 male lots and 59,258 female lots were bought in.

Table 3: Summary statistics for men and women: primary market sample

| Variables | Men |  |  |  | Women |  |  |  |
| :--- | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
|  |  | N | mean | sd |  | N | mean | sd |
| Auction participation | $4,180(4,050)$ | $0.969^{* * *}$ | 0.174 |  | $574(534)$ | $0.930^{* * *}$ | 0.255 |  |
| Total sales value (in $\$)$ | 4,050 | $3,381,389$ | $41,400,000$ |  | 534 | $1,536,746$ | $8,015,190$ |  |
| Year of birth | 4,180 | 1955 | 15.622 |  | 574 | 1958 | 14.990 |  |

The primary market sample consists of Western, contemporary artists only.
***The difference in proportions of the auction participation rates between men and women is statistically significant on a $1 \%$ significance level.
All prices are in constant $2017 \$$.

Table 4: Artist level regression results - Heckman selection model (primary market)

| Variables | Auction participation |  | Log of sales value |
| :--- | :---: | :---: | :---: |
|  | Selection model | Heckman model |  |
| Female | $-0.022^{* * *}$ |  | $-0.361^{* * *}$ |
|  | $(0.006)$ |  | $(0.112)$ |
| Year of birth | $-0.001^{* * *}$ |  | $-0.036^{* * *}$ |
|  | $(0.000)$ |  | $(0.002)$ |
| Artist Nationality Effects | Yes | Yes |  |
| Gallery Effects | Yes | Yes |  |
| Artist Effects | No | No |  |
| Observations | 4,754 | 4,548 |  |
| $\rho$ |  | $0.936^{* * *}$ |  |
|  |  | $(0.016)$ |  |
| Lambda |  | $2.310^{* * *}$ |  |
|  |  | $(0.055)$ |  |

Standard errors in parentheses. p<0.01, ${ }^{* *} \mathrm{p}<0.05$, $^{*} \mathrm{p}<0.1$.
The selection model shows the marginal effects at the mean.
The primary market sample consists of Western, contemporary artists only.

Table 5: Artwork level OLS regression results

| Variables | Log of real price |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Real price | Nominal price | Old Masters | Modern | Post War | Contemporary |
| Female | 0.044*** | $0.044^{* * *}$ | $0.100^{* * *}$ | 0.045*** | 0.149*** | -0.083*** |
|  | (0.004) | (0.004) | (0.011) | (0.007) | (0.008) | (0.007) |
| Design | -0.219*** | -0.219*** | -0.012 | -0.199*** | -0.261*** | -0.168*** |
|  | (0.003) | (0.003) | (0.009) | (0.006) | (0.006) | (0.009) |
| Photographs | -0.688*** | -0.688*** | $-0.707^{* * *}$ | -0.788*** | -0.718*** | -0.494*** |
|  | (0.004) | (0.004) | (0.014) | (0.007) | (0.008) | (0.007) |
| Prints \& multiples | -0.918*** | -0.918*** | $-0.897^{* * *}$ | $-1.017^{* * *}$ | -0.962*** | -0.804*** |
|  | (0.002) | (0.002) | (0.006) | (0.004) | (0.005) | (0.006) |
| Sculpture | $0.330^{* *}$ | 0.330*** | $0.322^{* * *}$ | $0.406 * * *$ | $0.341^{* * *}$ | $0.393 * * *$ |
|  | (0.003) | (0.003) | (0.008) | (0.007) | (0.007) | (0.007) |
| Works on paper | -0.409*** | -0.409*** | -0.379*** | -0.383*** | -0.371*** | -0.325*** |
|  | (0.002) | (0.002) | (0.005) | (0.004) | (0.005) | (0.006) |
| Eastern Europe | $0.014^{* * *}$ | 0.014*** | 0.441*** | 0.168*** | -0.528*** | -0.359*** |
|  | (0.005) | (0.005) | (0.010) | (0.007) | (0.011) | (0.012) |
| Northern Europe | $-0.272^{* * *}$ | $-0.272^{* * *}$ | $-0.228^{* * *}$ | $-0.130^{* * *}$ | $-0.497^{* * *}$ | $-0.057^{* * *}$ |
|  | (0.003) | (0.003) | (0.008) | (0.006) | (0.006) | (0.007) |
| Southern Europe | 0.149*** | 0.149*** | $0.107^{* * *}$ | 0.539*** | -0.228*** | -0.085*** |
|  | (0.003) | (0.003) | (0.010) | (0.006) | (0.007) | (0.008) |
| Western Europe | $-0.043^{* * *}$ | -0.043*** | 0.010 | 0.120*** | $-0.284^{* * *}$ | -0.100*** |
|  | (0.003) | (0.003) | (0.006) | (0.005) | (0.005) | (0.006) |
| Alive | -0.381*** | -0.381*** |  |  | -0.370*** |  |
|  | (0.002) | (0.002) |  |  | (0.004) |  |
| Log of size | 0.181*** | 0.181*** | 0.186*** | 0.144*** | 0.188*** | 0.240*** |
|  | (0.001) | (0.001) | (0.001) | (0.001) | (0.001) | (0.001) |
| Constant | $5.807^{* * *}$ | $5.454^{* * *}$ | 6.086*** | $5.820 * * *$ | $5.821^{* *}$ | $5.006^{* * *}$ |
|  | (0.035) | (0.035) | (0.078) | (0.049) | (0.077) | (0.117) |
| Observations | 2,677,190 | 2,677,190 | 539,186 | 854,843 | 556,518 | 418,504 |
| R-squared | 0.422 | 0.419 | 0.420 | 0.417 | 0.437 | 0.483 |
| Year Effects | Yes | Yes | Yes | Yes | Yes | Yes |
| Season Effects | Yes | Yes | Yes | Yes | Yes | Yes |
| Auction house Effects | Yes | Yes | Yes | Yes | Yes | Yes |

Table 6: Artwork level OLS regression results - by time period

| Variables | Log of real price |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
|  | 2000-2004 | 2005-2009 | 2010-2014 | 2015-2017 |
| Female | $0.074^{* * *}$ | 0.069*** | 0.019*** | 0.039** |
|  | (0.009) | (0.007) | (0.006) | (0.013) |
| Design | $-0.425^{* * *}$ | -0.207*** | -0.200*** | -0.186*** |
|  | (0.021) | (0.007) | (0.005) | (0.011) |
| Photographs | -0.945*** | $-0.722^{* * *}$ | $-0.563^{* * *}$ | -0.570*** |
|  | (0.009) | (0.007) | (0.006) | (0.014) |
| Prints \& multiples | $-1.175^{* * *}$ | -1.074*** | -0.736*** | -0.661*** |
|  | (0.005) | (0.004) | (0.004) | (0.009) |
| Sculpture | 0.406*** | 0.444*** | 0.289*** | 0.190*** |
|  | (0.008) | (0.007) | (0.005) | (0.011) |
| Works on paper | -0.439*** | -0.448*** | -0.378*** | -0.346*** |
|  | (0.005) | (0.004) | (0.004) | (0.008) |
| Eastern Europe | -0.016 | $0.140 * * *$ | 0.000 | -0.138*** |
|  | (0.012) | (0.008) | (0.007) | (0.015) |
| Northern Europe | $-0.260 * * *$ | -0.264*** | -0.247*** | -0.273*** |
|  | (0.007) | (0.006) | (0.005) | (0.011) |
| Southern Europe | 0.242*** | 0.188*** | $0.095^{* * *}$ | $0.083^{* * *}$ |
|  | (0.007) | (0.006) | (0.005) | (0.011) |
| Western Europe | 0.012** | -0.049*** | $-0.048^{* * *}$ | $-0.084^{* * *}$ |
|  | (0.006) | (0.005) | (0.004) | (0.009) |
| Alive | -0.487*** | -0.397*** | -0.357*** | -0.279*** |
|  | (0.005) | (0.004) | (0.003) | (0.007) |
| Log of size | 0.201*** | 0.204*** | $0.167^{* * *}$ | 0.147*** |
|  | (0.001) | (0.001) | (0.001) | (0.002) |
| Constant | $6.339 * * *$ | $6.596^{* * *}$ | $6.068^{* * *}$ | $6.321^{* * *}$ |
|  | (0.028) | (0.028) | (0.039) | (0.098) |
| Observations | 496,923 | 756,668 | 1,026,029 | 209,830 |
| R-squared | 0.452 | 0.443 | 0.424 | 0.402 |
| Year Effects | Yes | Yes | Yes | Yes |
| Season Effects | Yes | Yes | Yes | Yes |
| Auction house Effects | Yes | Yes | Yes | Yes |
| Standard errors in parentheses. ${ }^{* * *} \mathrm{p}<0.01,{ }^{* *} \mathrm{p}<0.05,{ }^{*} \mathrm{p}<0.1$. The base category for the object type is paintings. The base category for the region is North America. |  |  |  |  |
|  |  |  |  |  |
|  |  |  |  |  |

Table 7: Artwork level OLS regression results - by time period (contemporary sample)

| Variables | Log of real price |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
|  | 2000-2004 | 2005-2009 | 2010-2014 | 2015-2017 |
| Female | -0.035* | -0.040*** | $-0.093^{* * *}$ | -0.126*** |
|  | (0.020) | (0.014) | (0.011) | (0.022) |
| Design | -0.504*** | $-0.135^{* * *}$ | -0.126*** | -0.194*** |
|  | (0.061) | (0.019) | (0.013) | (0.028) |
| Photographs | $-0.583^{* * *}$ | $-0.514^{* * *}$ | -0.435*** | -0.520*** |
|  | (0.018) | (0.013) | (0.010) | (0.023) |
| Prints \& multiples | -1.061*** | $-0.993 * * *$ | -0.659*** | -0.654*** |
|  | (0.015) | (0.011) | (0.009) | (0.018) |
| Sculpture | 0.488*** | 0.471*** | 0.399*** | 0.271*** |
|  | (0.021) | (0.015) | (0.011) | (0.022) |
| Works on paper | -0.249*** | -0.313*** | -0.309*** | -0.375*** |
|  | (0.017) | (0.012) | (0.009) | (0.018) |
| Eastern Europe | $-0.286^{* * *}$ | $-0.221^{* * *}$ | $-0.384^{* * *}$ | -0.489*** |
|  | (0.040) | (0.023) | (0.017) | (0.033) |
| Northern Europe | -0.091*** | 0.056*** | -0.093*** | -0.102*** |
|  | (0.019) | (0.014) | (0.010) | (0.021) |
| Southern Europe | 0.050** | 0.033** | -0.211*** | -0.120*** |
|  | (0.024) | (0.016) | (0.012) | (0.025) |
| Western Europe | 0.034** | $-0.055^{* * *}$ | -0.165*** | -0.141*** |
|  | (0.017) | (0.012) | (0.009) | $(0.019)$ |
| Log of size | $0.257^{* * *}$ | 0.265*** | 0.234*** | 0.205*** |
|  | (0.004) | (0.003) | (0.002) | (0.004) |
| Constant | $5.442^{* * *}$ | 5.318*** | $5.454^{* * *}$ | 5.801*** |
|  | (0.144) | (0.082) | (0.138) | (0.254) |
| Observations | 44,731 | 106,980 | 185,573 | 43,267 |
| R-squared | 0.538 | 0.496 | 0.498 | 0.443 |
| Year Effects | Yes | Yes | Yes | Yes |
| Season Effects | Yes | Yes | Yes | Yes |
| Auction house Effects | Yes | Yes | Yes | Yes |
| Standard errors in parentheses. ${ }^{* * *} \mathrm{p}<0.01,{ }^{* *} \mathrm{p}<0.05$, $^{*} \mathrm{p}<0.1$. The base category for the object type is paintings. The base category for the region is North America. |  |  |  |  |
|  |  |  |  |  |
|  |  |  |  |  |

Table 8: Concentration of lots for male and female artists

| Share of lots | Men |  |  | Women |  |
| :--- | ---: | ---: | ---: | ---: | ---: |
|  | Share of artists | Share of value |  | Share of artists | Share of value |
| $25 \%$ | $0.30 \%(336)$ | $51.69 \%$ |  | $0.37 \%(21)$ | $23.62 \%$ |
| $50 \%$ | $2.09 \%(2,316)$ | $78.40 \%$ |  | $1.66 \%(93)$ | $65.65 \%$ |
| $75 \%$ | $8.87 \%(9,838)$ | $93.39 \%$ |  | $6.56 \%(368)$ | $93.36 \%$ |
| $90 \%$ | $23.96 \%(26,585)$ | $97.99 \%$ |  | $20.31 \%(1,140)$ | $98.77 \%$ |
| $95 \%$ | $39.95 \%(43,318)$ | $98.99 \%$ |  | $36.72 \%(2,061)$ | $99.54 \%$ |
|  | 110,938 | $117,335,262,644$ |  | 5,612 | $4,095,761,313$ |

All prices are in constant $2017 \$$.

Table 9: Weighted least squares regression results

| Variables | Log of real price |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Real price | Nominal price | Old Masters | Modern | Post War | Contemporary |
| Female | $-0.100^{* * *}$ | $-0.100^{* * *}$ | 0.049* | -0.010 | $-0.131^{* * *}$ | -0.042*** |
|  | (0.010) | (0.010) | (0.029) | (0.021) | (0.020) | (0.016) |
| Design | 0.229*** | 0.229*** | 0.500*** | $0.458^{* * *}$ | 0.230*** | 0.318*** |
|  | (0.011) | (0.011) | (0.034) | (0.023) | (0.019) | (0.023) |
| Photographs | -0.376*** | -0.376*** | -0.183*** | -0.360*** | -0.347*** | $-0.202^{* * *}$ |
|  | (0.011) | (0.011) | (0.041) | (0.029) | (0.022) | (0.015) |
| Prints \& multiples | -0.672*** | -0.672*** | -0.758*** | -0.522*** | -0.564*** | $-0.503^{* * *}$ |
|  | (0.009) | (0.009) | (0.026) | (0.017) | (0.019) | (0.020) |
| Sculpture | $0.437^{* * *}$ | 0.437*** | 0.649*** | 0.604*** | 0.348*** | 0.513*** |
|  | (0.010) | (0.010) | (0.023) | (0.021) | (0.018) | (0.016) |
| Works on paper | -0.342 ${ }^{* * *}$ | -0.342*** | -0.392*** | -0.219*** | -0.271*** | -0.195*** |
|  | (0.006) | (0.006) | (0.012) | (0.011) | (0.013) | (0.013) |
| Eastern Europe | $0.247^{* * *}$ | $0.247^{* * *}$ | 0.560*** | 0.290*** | 0.260*** | -0.013 |
|  | (0.012) | (0.012) | (0.028) | (0.023) | (0.023) | (0.022) |
| Northern Europe | -0.003 | -0.003 | -0.104*** | -0.116*** | 0.072*** | 0.002 |
|  | (0.009) | (0.009) | (0.019) | (0.023) | (0.016) | (0.019) |
| Southern Europe | 0.493*** | 0.493*** | 0.405*** | 0.194*** | 0.541*** | 0.084*** |
|  | (0.011) | (0.011) | (0.024) | (0.025) | (0.021) | (0.021) |
| Western Europe | $0.153^{* * *}$ | $0.153^{* * *}$ | -0.023 | -0.063*** | 0.265*** | -0.038** |
|  | (0.008) | (0.008) | (0.017) | (0.017) | (0.016) | (0.017) |
| Alive | -0.359*** | $-0.359^{* * *}$ |  |  | -0.132*** |  |
|  | (0.005) | (0.005) |  |  | (0.009) |  |
| Log of size | 0.138*** | 0.138*** | 0.202*** | 0.133*** | 0.095*** | 0.175*** |
|  | (0.002) | (0.002) | (0.004) | (0.004) | (0.003) | (0.003) |
| Constant | $5.607^{* * *}$ | $5.254^{* * *}$ | $5.424^{* * *}$ | 5.460*** | 5.794*** | 5.098*** |
|  | (0.040) | (0.040) | (0.088) | (0.060) | (0.061) | (0.138) |
| Observations | 2,677,190 | 2,677,190 | 539,186 | 854,843 | 556,518 | 418,504 |
| R-squared | 0.362 | 0.354 | 0.378 | 0.316 | 0.332 | 0.356 |
| Year Effects | Yes | Yes | Yes | Yes | Yes | Yes |
| Season Effects | Yes | Yes | Yes | Yes | Yes | Yes |
| Auction house Effects | Yes | Yes | Yes | Yes | Yes | Yes |

The base category for the object type is paintings.
The base category for the region is North America.
The weights are based on the number of observations
per artist.

Table 10: Quantiles for men and women

| Quantile | Total sales value (\$) | Men |  | Women |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | N artists | Cumulative | N artists | Cumulative |
| >99.97\% |  | 0.03\% (40) | 0.03\% | 0.00\% (0) | 0.00\% |
| <99.97\% | 452,388,320 | 0.01\%(17) | 0.05\% | 0.02\% (1) | 0.02\% |
| <99.96\% | 351,808,064 | 0.04\% (43) | 0.08\% | 0.05\% (3) | 0.07\% |
| <99.1\% | 176,461,520 | 0.90\% (994) | 0.98\% | 1.19\% (67) | 1.27\% |
| <99\% | 9,461,848 | 4.05\% (4,490) | 5.03\% | 3.06\% (172) | 4.33\% |
| <95\% | 982,622 | 5.09\% (5,650) | 10.12\% | 3.17\% (178) | 7.50\% |
| <90\% | 312,493 | 15.24\% (16,908) | 25.36\% | 10.23\% (574) | 17.73\% |
| < $75 \%$ | 50,209 | 25.19\% ( 27,949 ) | 50.56\% | 21.19\% (1,189) | 38.92\% |
| <50\% | 8,604 | 24.92\% ( 27,644 ) | 75.48\% | 26.60\% (1,493) | 65.52\% |
| <25\% | 2,089 | 14.82\% (16,442) | 90.30\% | 18.55\% (1,041) | 84.07\% |
| <10\% | 814 | 4.90\% (5,435) | 95.20\% | 6.99\% (392) | 91.05\% |
| <5\% | 545 | 4.80\% ( 5,326 ) | 100.00\% | 8.95\% (502) | 100.00\% |
| Total sales value | 121,431,023,957 |  |  |  |  |

Table 11: Group-specific quantiles for men and women

|  | Men |  |  | Women |  |
| :--- | :---: | :---: | :---: | :---: | :---: |
| Quantile | Total sales value (\$) | N artists |  | Total sales value (\$) | N artists |
| $>99.91 \%$ |  | 99 |  | 5 |  |
| $<99.91 \%$ | $176,750,048$ | 1,010 |  | $135,153,952$ | 51 |
| $<99 \%$ | $9,342,266$ | 4,437 |  | $12,382,016$ | 224 |
| $<95 \%$ | 992,138 | 5,547 |  | 724,759 | 281 |
| $<90 \%$ | 318,364 | 16,641 |  | 168,534 | 842 |
| $<75 \%$ | 51,854 | 27,735 |  | 23,281 | 1,403 |
| $<50 \%$ | 8,801 | 27,734 |  | 4,557 | 1,403 |
| $<25 \%$ | 2,147 | 16,640 |  | 1,288 | 841 |
| $<10 \%$ | 831 | 5,546 |  | 581 | 281 |
| $<5 \%$ | 554 | 5,549 |  | 403 | 281 |
| Total sales value | $117,335,262,644$ |  | $4,095,761,313$ |  |  |

Table 12: Quantile regression results

| Variables | Log of real price |  |  |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
|  | q25 | q 50 | q 75 | q 95 | q 99 | q 99.9 |
| Female | $0.045^{* * *}$ | $0.049^{* * *}$ | $0.054^{* * *}$ | $0.067^{* * *}$ | $-0.018^{* * *}$ | $-0.091^{* * *}$ |
| Auction House Effects | $(0.000)$ | $(0.000)$ | $(0.000)$ | $(0.000)$ | $(0.000)$ | $(0.000)$ |
| Region Effects | Yes | Yes | Yes | Yes | Yes | Yes |
| Alive Dummy | Yes | Yes | Yes | Yes | Yes | Yes |
| Size Effects | Yes | Yes | Yes | Yes | Yes | Yes |
| Artist Effects | Yes | Yes | Yes | Yes | Yes | Yes |
| Observations | No | No | No | No | No | No |
| P-values based on bootstrapped standard errors in parentheses. ${ }^{* * *} \mathrm{p}<0.01,{ }^{* *} \mathrm{p}<0.05,{ }^{*} \mathrm{p}<0.1$. |  |  |  |  |  |  |

## Appendix


(a) Index by gender - Full sample

(b) Index by genders - Contemporary sample

Figure A1: Price index by gender


Figure A2: Evolution female-to-male ratio

Table A1: Top 25 male artists by value of sales

| Rank | Artist | Origin | Total sales <br> value in $\$$ <br> (market share) | Total sales <br> volume <br> (market share) | Buy-in <br> rate |
| :--- | :--- | :--- | ---: | ---: | ---: |
| 1 | Pablo Picasso | Southern Europe | $5,853,551,616(4.99 \%)$ | $37,386(1.45 \%)$ | 0.247 |
| 2 | Andy Warhol | North America | $4,931,258,880(4.20 \%)$ | $19,028(0.74 \%)$ | 0.310 |
| 3 | Claude Monet | Western Europe | $2,509,770,496(2.14 \%)$ | $493(0.02 \%)$ | 0.197 |
| 4 | Gerhard Richter | Western Europe | $2,128,574,336(1.81 \%)$ | $3,587(0.14 \%)$ | 0.255 |
| 5 | Francis Bacon | Northern Europe | $2,071,435,648(1.77 \%)$ | $1,372(0.05 \%)$ | 0.235 |
| 6 | Alberto Giacometti | Western Europe | $1,661,223,808(1.42 \%)$ | $1,991(0.08 \%)$ | 0.309 |
| 7 | Jean-Michel Basquiat | North America | $1,604,688,384(1.37 \%)$ | $1,308(0.05 \%)$ | 0.288 |
| 8 | Mark Rothko | North America | $1,589,495,040(1.35 \%)$ | $142(0.01 \%)$ | 0.184 |
| 9 | Henri Matisse | Western Europe | $1,384,500,224(1.18 \%)$ | $5,157(0.20 \%)$ | 0.302 |
| 10 | Roy Lichtenstein | North America | $1,365,195,904(1.16 \%)$ | $6,429(0.02 \%)$ | 0.247 |
| 11 | Amedeo Modigliani | Southern Europe | $1,282,909,952(1.09 \%)$ | $502(0.58 \%)$ | 0.344 |
| 12 | Marc Chagall | Western Europe | $1,246,740,480(1.06 \%)$ | $14,957(0.57 \%)$ | 0.294 |
| 13 | Joan Miró | Southern Europe | $1,195,891,584(1.02 \%)$ | $14,781(0.21 \%)$ | 0.285 |
| 14 | Willem De Kooning | North America | $1,144,317,696(0.98 \%)$ | $1,272(0.06 \%)$ | 0.272 |
| 15 | Lucio Fontana | Southern Europe | $1,098,615,296(0.94 \%)$ | $2,772(0.11 \%)$ | 0.266 |
| 16 | Alexander Calder | North America | $1,088,666,752(0.93 \%)$ | $5,479(0.05 \%)$ | 0.238 |
| 17 | Pierre-Auguste Renoir | Western Europe | $1,046,396,352(0.89 \%)$ | $3,766(0.15 \%)$ | 0.309 |
| 18 | Zao Wou-Ki | Western Europe | $1,015,000,512(0.87 \%)$ | $4,045(0.15 \%)$ | 0.206 |
| 19 | Fernand Léger | Western Europe | $1,005,042,112(0.86 \%)$ | $2,978(0.16 \%)$ | 0.354 |
| 20 | Cy Twombly | North America | $850,141,376(0.72 \%)$ | $881(0.06 \%)$ | 0.765 |
| 21 | Jeff Koons | North America | $848,892,096(0.72 \%)$ | $1,646(0.12 \%)$ | 0.296 |
| 22 | Paul Cézanne | Western Europe | $791,902,080(0.67 \%)$ | $697(0.05 \%)$ | 0.299 |
| 23 | Edgar Degas | Western Europe | $771,783,232(0.66 \%)$ | $1,274(0.17 \%)$ | 0.295 |
| 24 | René Magritte | Western Europe | $734,759,296(0.63 \%)$ | $1,519(0.03 \%)$ | 0.235 |
| 25 | Damien Hirst | Northern Europe | $705,134,592(0.60 \%)$ | $3,940(0.03 \%)$ | 0.406 |

All prices are in constant $2017 \$$.

Table A2: Top 25 female artists by value of sales

| Rank | Artist | Origin | Total sales value in $\$$ (market share) | Total sales volume (market share) | $\begin{array}{r} \text { Buy-in } \\ \text { rate } \end{array}$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | Joan Mitchell | North America | 392,962,816 (9.59\%) | 641 (0.61\%) | 0.213 |
| 2 | Georgia O'Keeffe | North America | 211,702,064 (5.17\%) | 117 (0.11\%) | 0.204 |
| 3 | Louise Bourgeois | North America | 197,968,512 (4.83\%) | 649 (0.62\%) | 0.289 |
| 4 | Agnes Martin | North America | 193,711,040 (4.73\%) | 296 (0.28\%) | 0.249 |
| 5 | Cindy Sherman | North America | 140,606,176 (3.43\%) | 1,269 (1.21\%) | 0.268 |
| 6 | Barbara Hepworth | Northern Europe | 135,153,952 (3.30\%) | 616 (0.59\%) | 0.146 |
| 7 | Tamara De Lempicka | Eastern Europe | 127,470,128 (3.11\%) | 313 (0.30\%) | 0.357 |
| 8 | Natalia Sergeevna Goncharova | Eastern Europe | 127,109,512 (3.10\%) | 731 (0.70\%) | 0.463 |
| 9 | Mary Cassatt | North America | 88,247,688 (2.15\%) | 832 (0.79\%) | 0.296 |
| 10 | Helen Frankenthaler | North America | 79,406,904 (1.94\%) | 1,100 (1.05\%) | 0.253 |
| 11 | Bridget Riley | Northern Europe | 78,610,368 (1.92\%) | 818 (0.78\%) | 0.189 |
| 12 | Berthe Morisot | Western Europe | 76,978,256 (1.88\%) | 258 (0.25\%) | 0.340 |
| 13 | Eileen Gray | Northern Europe | 75,399,800 (1.84\%) | 184 (0.18\%) | 0.326 |
| 14 | Gabriele Münter | Western Europe | 67,722,952 (1.65\%) | 449 (0.43\%) | 0.231 |
| 15 | Niki De Saint Phalle | Western Europe | 67,633,304 (1.65\%) | 1,849 (1.76\%) | 0.361 |
| 16 | Maria Helena Vieira Da Silva | Western Europe | 62,461,532 (1.53\%) | 683 (0.65\%) | 0.320 |
| 17 | Elisabeth Frink | Western Europe | 56,816,528 (1.39\%) | 1,212 (1.16\%) | 0.186 |
| 18 | Camille Claudel | Western Europe | 47,351,292 (1.16\%) | 115 (0.11\%) | 0.275 |
| 19 | Julie Mehretu | North America | 39,050,448 (0.95\%) | 117 (0.11\%) | 0.328 |
| 20 | Marie Laurencin | Western Europe | 37,916,940 (10.93\%) | 1,633 (1.56\%) | 0.452 |
| 21 | Germaine Richier | Western Europe | 36,489,668 (0.89\%) | 207 (0.20\%) | 0.310 |
| 22 | Charlotte Perriand | North America | 36,297,372 (0.89\%) | 1,270 (1.21\%) | 0.367 |
| 23 | Sonia Delaunay | Western Europe | 35,823,440 (0.87\%) | 2,414 (0.23\%) | 0.412 |
| 24 | Zinaida Evgenievna Serebryakova | Eastern Europe | 46,413,028 (0.87\%) | 130 (0.12\%) | 0.272 |
| 25 | Elizabeth Peyton | North America | 34,532,152 (0.84\%) | 305 (0.29\%) | 0.343 |

All prices are in constant $2017 \$$.
Table A3: Artwork level OLS regression results - by generation of artist

| Variables | Log of real price |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | <1700 | <1800 | <1825 | <1850 | <1875 | <1900 | <1925 | <1950 | <1975 | $<2001$ |
| Female | $0.358^{* * *}$ | 0.058 | $0.125^{* * *}$ | $0.340 * * *$ | 0.008 | -0.008 | $0.124^{* * *}$ | 0.080*** | -0.105*** | $-0.097^{* * *}$ |
|  | (0.069) | (0.041) | (0.031) | (0.024) | (0.013) | (0.008) | (0.008) | (0.008) | (0.011) | (0.033) |
| Design | -0.668*** | -0.292*** | 0.045 | $0.049^{* * *}$ | $-0.012$ | $-0.171^{* * *}$ | -0.248*** | -0.312*** | -0.068*** | 0.180*** |
|  | (0.092) | (0.048) | (0.041) | (0.017) | (0.011) | (0.007) | (0.007) | (0.009) | (0.016) | (0.055) |
| Photographs | -0.708* | -0.469*** | $-0.231^{* * *}$ | $-0.775^{* * *}$ | $-0.689^{* * *}$ | $-0.787^{* * *}$ | -0.658*** | -0.756*** | -0.406*** | $-0.461 * * *$ |
|  | (0.371) | $(0.105)$ | $(0.020)$ | (0.026) | $(0.017)$ | $(0.010)$ | (0.008) | (0.008) | (0.010) | (0.037) |
| Prints and multiples | $-1.267^{* * *}$ | $-1.441^{* * *}$ | $-1.154^{* * *}$ | $-0.999^{* * *}$ | $-0.851^{* * *}$ | $-1.003^{* * *}$ | $-1.010^{* * *}$ | $-1.025^{* * *}$ | $-0.742^{* * *}$ | $-0.738^{* * *}$ |
|  | (0.013) | $(0.012)$ | $(0.019)$ | $(0.014)$ | (0.008) | $(0.005)$ | (0.005) | (0.005) | (0.010) | (0.042) |
| Sculpture | $0.225^{* * *}$ | $0.061^{* * *}$ | $0.118^{* * *}$ | $0.352^{* * *}$ | $0.354^{* * *}$ | $0.400^{* * *}$ | $0.452^{* * *}$ | $0.314^{* * *}$ | $0.483^{* * *}$ | $0.103^{* *}$ |
|  | $(0.026)$ | $(0.019)$ | $(0.017)$ | $(0.014)$ | (0.011) | (0.008) | (0.008) | (0.007) | (0.012) | (0.042) |
| Works on paper | $-0.601^{* * *}$ | $-0.467^{* * *}$ | $-0.402^{* * *}$ | $-0.475^{* * *}$ | $-0.348^{* * *}$ | $-0.390^{* * *}$ | $-0.355^{* * *}$ | $-0.367^{* * *}$ | $-0.248^{* * *}$ | -0.326 ${ }^{* * *}$ |
|  | $(0.013)$ | (0.009) | $(0.010)$ | (0.009) | $(0.006)$ | $(0.005)$ | $(0.005)$ | $(0.006)$ | (0.010) | (0.038) |
| Eastern Europe | 0.392* | $0.119^{* *}$ | $0.567^{* * *}$ | 0.513*** | $0.427^{* * *}$ | $0.134^{* * *}$ | -0.182*** | -0.499*** | -0.431*** | $-0.399^{* * *}$ |
|  | (0.201) | (0.049) | (0.031) | (0.020) | (0.012) | (0.009) | (0.010) | (0.012) | (0.019) | (0.046) |
| Northern Europe | 0.060 | -0.537*** | -0.659*** | $-0.209^{* * *}$ | $-0.224^{* * *}$ | $-0.145^{* * *}$ | -0.221*** | -0.476*** | 0.058*** | $-0.197^{* * *}$ |
|  | (0.189) | (0.020) | (0.018) | (0.013) | (0.010) | (0.008) | (0.007) | (0.007) | (0.010) | (0.039) |
| Southern Europe | $0.491 * * *$ | -0.085*** | -0.183*** | 0.174*** | 0.089*** | $0.607^{* * *}$ | $0.027^{* * *}$ | -0.294*** | -0.168*** | $-0.300^{* * *}$ |
|  | (0.188) | (0.021) | (0.027) | (0.018) | (0.013) | (0.007) | (0.007) | (0.008) | (0.014) | (0.059) |
| Western Europe | $0.500^{* * *}$ | -0.299*** | -0.327*** | 0.055*** | -0.008 | 0.081*** | -0.022*** | -0.335*** | -0.212*** | $-0.322^{* * *}$ |
|  | (0.188) | (0.020) | (0.017) | (0.011) | (0.008) | (0.006) | (0.006) | (0.006) | (0.010) | (0.038) |
| Alive |  |  |  |  |  |  | $-0.079^{* * *}$ $(0.00 .5)$ | $-0.322^{* * *}$ <br> (0.004) | $-0.659^{* * *}$ |  |
|  |  |  |  |  |  |  | (0.005) | (0.004) | (0.013) |  |
| Log of size | $0.133^{* * *}$ | $0.224^{* * *}$ | 0.229*** | $0.206^{* * *}$ | 0.182*** | $0.147^{* * *}$ | 0.182*** | $0.222^{* * *}$ | $0.287^{* * *}$ | 0.220*** |
|  | (0.003) | (0.002) | (0.003) | (0.003) | (0.002) | (0.001) | (0.001) | (0.001) | (0.002) | (0.008) |
| Constant | 7.429*** | $7.457^{* * *}$ | $5.993 * * *$ | $6.197^{* * *}$ | $6.049^{* * *}$ | $5.887^{* * *}$ | 5.560*** | $5.545^{* * *}$ | $5.193 * * *$ | 5.409*** |
|  | (0.207) | (0.075) | (0.502) | (0.187) | (0.086) | (0.058) | (0.073) | (0.081) | (0.278) | (0.688) |
| Observations | 125,023 | 103,448 | 103,135 | 171,749 | 343,970 | 612,586 | 521,910 | 437,473 | 159,572 | 11,237 |
| R-squared | 0.455 | 0.440 | 0.415 | 0.445 | 0.413 | 0.419 | 0.412 | 0.481 | 0.533 | 0.589 |
| Year Effects | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes |
| Season Effects | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes |
| Auction house Effects | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes | Yes |

Table A4: Artwork level OLS regression results - price scaled by estimate

| Variables | Ratio nominal price to action house pre-sale estimate |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | Pooled | Old Masters | Modern | Post War | Contemporary |
| Female | -0.667 | -0.085 | 0.458 | -1.777 | -2.268 |
|  | (1.070) | (2.116) | (1.932) | (2.453) | (2.764) |
| Design | $2.915^{* * *}$ | -0.353 | 4.617*** | 5.109*** | -4.058 |
|  | (0.907) | (1.670) | (1.597) | (1.956) | (3.531) |
| Photographs | 0.043 | -0.151 | 0.794 | -0.783 | -0.505 |
|  | (1.036) | (2.573) | (2.081) | (2.373) | (2.538) |
| Prints \& multiples | 0.905 | -0.348 | 2.055* | 0.157 | -0.278 |
|  | (0.654) | (1.278) | (1.203) | (1.572) | (2.262) |
| Sculpture Sculpture | 0.952 | 0.229 | 1.787 | -1.092 | 1.279 |
|  | (0.935) | (1.610) | (1.963) | (2.051) | (2.755) |
| Works on paper | 0.212 | -0.160 | 0.616 | -0.473 | -0.600 |
|  | (0.609) | (0.917) | (1.111) | (1.601) | (2.285) |
| Eastern Europe | 1.383 | -0.131 | 2.823 | -0.712 | -0.697 |
|  | (1.235) | (1.891) | (2.037) | (3.311) | (4.501) |
| Northern Europe | 0.084 | 0.448 | -4.825*** | 1.680 | 3.668 |
|  | (0.841) | (1.408) | (1.753) | (1.919) | (2.641) |
| Southern Europe | 0.055 | 0.040 | -0.875 | -1.158 | -0.712 |
|  | (0.880) | (1.939) | (1.637) | (2.068) | (3.093) |
| Western Europe | 0.439 | 0.011 | 0.887 | -0.724 | -1.083 |
|  | (0.686) | (1.185) | (1.329) | (1.613) | (2.345) |
| Alive | 1.060** |  |  | -1.356 |  |
|  | (0.529) |  |  | (1.162) |  |
| Log of Size | -0.232 | 0.086 | 0.035 | -1.200*** | -0.335 |
|  | (0.148) | (0.265) | (0.289) | (0.350) | (0.507) |
| Constant | -0.055 | -0.866 | -3.362 | 11.674 | 0.570 |
|  | (9.119) | (14.288) | (13.272) | (22.828) | (44.033) |
| Observations | 2,434,732 | 479,566 | 772,853 | 515,204 | 392,850 |
| R-squared | 0.001 | 0.000 | 0.003 | 0.001 | 0.004 |
| Year Effects | Yes | Yes | Yes | Yes | Yes |
| Season Effects | Yes | Yes | Yes | Yes | Yes |
| Auction house Effects | Yes | Yes | Yes | Yes | Yes |
| Standard errors in parentheses. ${ }^{* * *} \mathrm{p}<0.01,{ }^{* *} \mathrm{p}<0.05,^{*} \mathrm{p}<0.1$. The base category for the object type is paintings. The base category for the region is North America. |  |  |  |  |  |
|  |  |  |  |  |  |
|  |  |  |  |  |  |

Table A5: Quantile regression on OLS residual

| Variables | OLS residual |  |  |  |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | q 5 | q 25 | q 50 | q 75 | q 95 | q 99 | q 99.9 |
| Female | $0.029^{* * *}$ | $0.033^{* * *}$ | $0.055^{* * *}$ | $0.096^{* * *}$ | $0.079^{* * *}$ | $-0.068^{* * *}$ | 0.031 |
|  | $(0.000)$ | $(0.000)$ | $(0.000)$ | $(0.000)$ | $(0.000)$ | $(0.000)$ | $(0.689)$ |
| Observations | $2,677,190$ | $2,677,190$ | $2,677,190$ | $2,677,190$ | $2,677,190$ | $2,677,190$ | $2,677,190$ |

P-values based on bootstrapped standard errors in parentheses. ${ }^{* * *} \mathrm{p}<0.01,{ }^{* *} \mathrm{p}<0.05,^{*} \mathrm{p}<0.1$. The residual is based on a linear regression of the object type, auction house, season, region, alive dummy, artists dummies and size are regressed on the logarithm of the artwork price .


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[^1]:    ${ }^{1}$ Lower status characteristics might refer for instance to age, education race or gender.

[^2]:    ${ }^{2}$ For instance better negotiation and self-promotion skills.

[^3]:    ${ }^{3}$ One approach to disentangle the supply from the demand side would be to perform the analysis separately for common names where there is not doubt about the gender (e.g Mary) and ambiguous names (e.g. Kim) where one would not have full certainty about the gender of the artist. However, it is not a plausible assumptions that auction houses or auction participants are uninformed about the identity of the artist. For instance, auction catalogs typically provide some information about the artist using pronouns which disclose the gender of the artist.

[^4]:    ${ }^{4}$ Decorative Art includes among other antiques, ceramics, furniture, jewelry, and watches.
    ${ }^{5}$ The employed consumer price index can be found under https://data.oecd.org/price/inflation-cpi.htm.

[^5]:    ${ }^{9}$ Nationality is defined on country level and includes all countries in Europe and North America totaling to 53 countries. Due to collinearity concerns, 5 of these nationalities were included in the regression model.

[^6]:    ${ }^{10}$ In order to avoid overparameterization, galleries that represented less than 100 artists were subsumed under the category 'others'. This resulted in 23 gallery dummies. Due to collinearity concerns, 9 of these galleries were included in the regression model.

[^7]:    ${ }^{11}$ In total, there are 1,522 auction houses in our data set. Due to collinearity concerns we subsumed auction houses below the 90th quantile in terms of number of transactions under "other". This resulted in 270 different categories.
    ${ }^{12}$ All countries are split into five regions: North America, Eastern Europe, Northern Europe, Southern Europe and Western Europe.

[^8]:    ${ }^{13}$ To provide further homogenize our sample, we also consider every cohort of artist separately and run regressions for each generation of artists whereby one generation is defined by a time period of 25 years. The results are presented in Table A3 in the Appendix. Consistent with the previous results, we observe a premium on female lots for the generations active before the year 1850 and a discount for more recent generations born after 1950 .

[^9]:    ${ }^{14}$ Before an auction takes place, auction houses typically publish a catalogue listing all lots that will be for sale with their own estimated value of these lots.

