**INTRODUCTION**

Imagine you are heading to Christie's to bid on a Monet painting, which experts believe is worth $1 million based on its size and medium. However, you’re unaware of those experts’ estimates and you have no idea how much the Monet is worth, so you look up recent sales of other Impressionist paintings. You find out a Van Gogh sold for $20 million last week. With that number in mind, you start to believe the Monet is also worth $20 million, and that’s how much you bid – even if that amount reflects more the price of the Van Gogh than the Monet itself.

This is known as the *anchoring effect* - a well-studied cognitive bias in which the first value you hear (the “anchor”) can shape your perception of what sorts of values are normal. This was first demonstrated in a landmark experiment by Tversky & Kahneman (1974)[[1]](#footnote-1). In that study, participants were given five seconds to mentally calculate the product of the numbers 1 through 8, visually written out for them on a blackboard. It was found that if the numbers were displayed as “1x2x3x4x5x6x7x8”, participants would read the lower numbers first and give a median estimate of 512 for the product. On the other hand, if the numbers were displayed as “8x7x6x5x4x3x2x1”, then participants would first see the higher numbers and give a much higher median estimate of 2250. Thus, it seems that first impressions do affect judgement, at least in quantitative scenarios.

This bias also appears in the market for fine art auctions, which in 2014 saw a sales volume of approximately $7.35 billion.[[2]](#footnote-2) To the best of our knowledge, Beggs & Graddy (2009) were the first to formally study anchoring in the context of art auctions, and describe it as follows. The hedonic value of an artwork, say a painting, is assumed to be determined by its hedonic characteristics such as the artist, medium, and degree of authenticity. These features do not change over time[[3]](#footnote-3), which means that if buyers were perfectly rational, they would pay only according to their (time-dependent) demand for these features. If however, buyers learn that the painting previously sold for a very high price, they may internalize that previous high price as a reference point – the “anchor” – and be willing to pay much more as a result. Past price could legitimately inform current price since it reflects hedonic value, as well as other relevant, non-hedonic influences such as bidder wealth and artwork reputation. However, if we control for those, then past price becomes an irrelevant signal in this context, and its impact on current price can be interpreted as an anchoring effect. Thus, anchoring is said to occur when past price biases current price, given hedonic value and non-hedonic price determinants are controlled for.

Of course, the ways in which auction participants internalize and act upon past price cannot be inferred from just observing prices. Hence in our research and in much of the literature, including Beggs & Graddy (2009), this is treated as a black box. The mere observation of this effect, i.e. past price biasing current price, suffices for our definition of anchoring (discussed further in Section <>).

Using a regression model that carefully separates out anchoring, Beggs & Graddy (2009) identified and analyzed resales of Impressionist and Contemporary paintings, finding significant evidence of anchoring effects. However, as mentioned in their paper, it is very difficult to identify multiple sales of the same art piece. This method of testing for anchoring effects cannot be applied to new works or works that have never been brought to auction. Moreover, in practice, it turns out that auction specialists appraise an art piece based primarily on past sales of related pieces[[4]](#footnote-4). The anchoring research of Beggs & Graddy (2009) is limited in its analysis and applicability.

In this paper, we study whether the sales of similar paintings (substitutes) display anchoring cross-effects – for example, whether the past price of a Monet can bias the current price of a Van Gogh. Our data consists of two datasets of Impressionist and Contemporary art, used not only in Beggs & Graddy (2009) but also in many other econometric papers on art auctions, as well as a new dataset of assorted art sales (2006-2015) specifically collected for this project. To show we understand the original regression model of Beggs & Graddy (2009), we begin by successfully replicating their general anchoring findings. Our success is surprising and noteworthy because, as discussed later in Section <>, we do not know exactly which observations in their data were originally used. Next, we introduce our expanded version of their model, which tests for anchoring cross-effects by controlling for similarity across pieces. We introduce two quantitative measures of similarity. Running our model on these three datasets, we discover significant evidence of anchoring cross-effects for Impressionist and assorted art. To experiment further, we also run our regressions on a subset of our assorted art dataset for three artist pairs: Joan Miro & Salvador Dali, Pablo Picasso & Marc Chagall, and Edvard Munch & Henri de Toulouse-Lautrec. We explain this selection of artists in Section <>. We find the strongest and most significant evidence of anchoring cross-effects between works by Picasso and Chagall.

This research makes several major contributions to the existing literature on art auctions. First, to the best of our knowledge, no econometric work has attempted to quantify hedonic similarity between art pieces. Thus, this study is useful not only for appraising art, but also for other tasks where art pieces must be compared, such as constructing price indices across heterogeneous artworks. We hope our two measures of similarity may provide a starting point for such analysis. Second, much of the econometric work on art auctions has relied on the same two Impressionist and Contemporary art datasets that only include sales until 1991 and 1994, respectively. Our new dataset of approximately 500,000 assorted art sales (2006-2015), constructed by writing a Python program to scrape Blouin ArtInfo for 10 straight days, is a larger and more up-to-date collection of auction data. Lastly, our discovery of anchoring cross-effects is notable because it adds to the growing body of research on how price signals implicitly propagate around the art auction market. Our work allows researchers to account for hidden biases such as anchoring when comparing sales of different art pieces, and demonstrates how Beggs & Graddy’s original anchoring model may be successfully customized. For auction houses and professionals, our work provides a practical regression model for estimating an artwork’s price in light of related sales. Beggs & Graddy’s original model has been extensively applied in other domains such as corporate finance[[5]](#footnote-5), real estate[[6]](#footnote-6), and horse racing[[7]](#footnote-7). Our approach is more general than theirs, and may find application in other fields as well.

As part of our research, we conducted interviews with several specialists and experts in the field. To learn about the art market and how auction specialists appraise pieces, we talked with Mark Best, a former financial analyst who is now a specialist in American, Modern, and Contemporary prints at Sotheby’s NYC. For insight into artistic similarity, we talked with Hadley Newton, who formerly worked at Sotheby’s with Impressionist art. The three artist pairs we examine in Section <> were suggested by Hadley. We also talked extensively with Raphaelle Benabou, who is an administrator of art collections, estates, and valuations at Bonham’s in London. Raphaelle provided us with many of our anecdotes and fact-checked our description of the auction system. We draw upon these interviews for both our discussion and our quantitative analysis.

Determining artistic similarity is not trivial: we were told that no two art pieces are the same[[8]](#footnote-8). Even in the case of prints, where 100-200 copies (editions) of the same art piece are manufactured and numbered in order of production, an edition with a lower number (i.e. produced earlier) may sell for more than an edition with a higher number. Furthermore, drivers of similarity can vary at different price points, and whether art is purchased as a decoration or as an investment. In this paper, we provide a starting point for quantitatively measuring similarity between pieces, but acknowledge that better measures could be constructed.

This thesis proceeds as follows. In Section I, we give a brief overview of the art auction system and process, followed by a discussion of anchoring and its role in this market. Section II surveys the literature relevant to anchoring in the art market and explains the importance of our research. Section III describes our methodology, which includes the original regressions of Beggs & Graddy (2009), our expanded regression models, and our measures of substitution. Section IV describes the original data of Beggs & Graddy, as well as our new dataset, and explains the motivation behind constructing the latter. Section V gives our results. This includes our replication of the anchoring work of Beggs & Graddy, followed by our findings on anchoring cross-effects. We then present the results of our experiments on the three artist pairs suggested to us by Hadley Newton. Finally, Section VII concludes with a summary of our research and directions for future work.

1. Tversky, Amos, and Daniel Kahneman. "Judgment under uncertainty: Heuristics and biases." *science* 185.4157 (1974): 1124-1131. [↑](#footnote-ref-1)
2. <http://www.christies.com/about/press-center/releases/pressrelease.aspx?pressreleaseid=7712>

   <http://www.xe.com/currencyconverter/convert/?From=GBP&To=USD> accessed 2/20/2015 [↑](#footnote-ref-2)
3. Age is not considered a hedonic feature, because for us it represents time-dependent demand for art rather than intrinsic characteristics of the work. [↑](#footnote-ref-3)
4. Interview with Raphaelle Benabou [↑](#footnote-ref-4)
5. Dougal, Casey, et al. "Anchoring on credit spreads." *The Journal of Finance*70.3 (2015): 1039-1080. [↑](#footnote-ref-5)
6. Leung, Tin Cheuk, and Kwok Ping Tsang. "Anchoring and loss aversion in the housing market: implications on price dynamics." *China Economic Review* 24 (2013): 42-54. [↑](#footnote-ref-6)
7. McAlvanah, Patrick, and Charles C. Moul. "The house doesn’t always win: Evidence of anchoring among Australian bookies." *Journal of Economic Behavior & Organization* 90 (2013): 87-99. [↑](#footnote-ref-7)
8. Interview, Mark Best. [↑](#footnote-ref-8)