# Left-Digit Bias, Investor Attention and Trading Behavior

John George Edika Neil Gathergood\* Loewenstein<sup>†</sup> Quispe–Torreblanca<sup>‡</sup> Stewart<sup>§</sup>

March 25, 2020

#### **Abstract**

Abstract here

Keywords: words
JEL Codes: codes

<sup>\*</sup>University of Nottingham, School of Economics; Network for Integrated Behavioural Science. Email: john.gathergood@nottingham.ac.uk.

<sup>&</sup>lt;sup>†</sup> Social and Decision Sciences, Carnegie Mellon University. Email: gl20@andrew.cmu.edu.

<sup>‡</sup> University of Oxford, Saïd Business School. Email: Edika.Quispe-Torreblance@sbs.ox.ac.uk.

<sup>§</sup> University of Warwick, Warwick Business School. Email: Neil.Stewart@wbs.ac.uk.

Although there is by now a large literature in finance addressing the question of when people *don't* like to sell stocks – specifically focusing on the disposition effect, the distaste for selling stocks at a nominal loss – beyond this strong regularity there is very little research focusing on when, exactly, investors *do* sell stocks. Are there specific events that trigger the sale of a stock? Recent research (Akepanidtaworn et al., 2019), which finds that the buy decisions of professional traders are quite sensible – the stocks they buy are more likely to rise in value than those they don't buy – but that their sell decisions are worse than random, further highlights the need for a better understanding of when stock sales occur.

While not providing a comprehensive theory, nor a broad empirical investigation, of when people sell stocks, in this paper we address one event that, we predicted and found, has a substantial effect on sales: People are significantly -xx% – more likely to sell stocks when their price crosses a round-number price threshold from below – e.g., rising from below \$30 per share to above \$30 per share. By the same token, we find that investors are less likely to sell stocks immediately after they cross a round number threshold from above. We document these interrelated patterns using a data set of transactions made by online retail investors, demonstrate its robustness across different empirical inspections, and rule out limit orders as an alternative explanation.

Left-digit bias is the tendency to focus on the leftmost digit of a number while paying less attention to other digits (Poltrock and Schwartz, 1984). Prior research on the left-digit bias has shown automobiles depreciate disproportionately when their milage crosses a around number threshold. Research on physician decision making likewise find that patients hospitalized with acute myocardial infarction 2 weeks after, as compared with 2 weeks before, their 80th birthday were significantly less likely to undergo coronary-artery bypass graft surgery. And research (Shlain, 2018) shows not only that 99 cent pricing works – that consumers respond to a one cent increase of \$.99 to \$1.00 as if it was a 15-25 cent difference, but also that firms exploit this bias less than they would if they were maximizing profits. Our contribution is to show that the left-digit bias strongly affects the behavior of investors.

## 1 Data

Data were provided by Barclays Stockbroking, an execution-online brokerage service operating in the United Kingdom. The data cover the period April 2012 to March 2016 and include daily-level records of trades and quarterly-level records of portfolio positions. The data also include a dummy variable, at daily frequency, denoting whether the investor made a login to their account on the day. The daily-level login dummy variable covers all days, including days on which the market is closed such as Sundays and public holidays, which we use later in our analysis. We combine the daily-level records of trades with the quarterly-level records of portfolio positions, together with stock price data from Datastream, to calculate the value of each stock position in an investor's portfolio on each day of the sample period.

## 1.1 Sample Selection

As a first step, we apply a series of data cleaning sample restrictions which restrict the data to active accounts with trading histories during the data period for which we can match price and demographic data. Details of this first stage of data cleaning are shown in Table 1. The unrestricted sample as received from Barclays contains 155,300 accounts. In this version of the paper we draw a 60% random sample of accounts for analysis.

The unit of observation in the data is an account  $\times$  stock  $\times$  day, i.e. an observation per investor per stock holding per day. We focus our analysis on two subsets of this universe of account  $\times$  days, specifically login-days and sell-days. We define a login-day as an observation which is paired with a login and a sell-day as an observation which is paired with a sale event on the day from the portfolio (of the stock, or of a different stock held in the account on the same day). The sample of accounts together provides a total of approximately 135 million login-days and approximately 100,000 sell-days.

We then apply six data cleaning restrictions to obtain a baseline sample, which are applied to the data at the account level unless otherwise noted. We apply these restrictions in order to limit the sample to the minimum variables required for analysis. First, we drop observations

<sup>&</sup>lt;sup>1</sup> During the data period the brokerage operated only through an online interface. Barclays have subsequently introduced a mobile phone trading app.

for which the account is inactive, defined as a one-year period in which the investor makes fewer than two logins or two transactions. Where an account does not meet this restriction, we drop all observations for the relevant year. Second, we remove observations where a matched price is not available from Datastream. Third, we remove observations for all account  $\times$  days in which there are fewer than two stocks within the portfolio. Fourth, we remove all observations for accounts for which demographic data is missing (i.e., we drop all investor  $\times$  stock  $\times$  days for that account from the sample). Fifth, we remove the days in which the investor purchased the stocks (starting position days) as speculative day traiding is rare among retail investors. Finally, we remove accounts with extreme portfolio values (99 percentile of average portfolio value).

Table 1 reports the effects of these steps in sample selection. The table reports the number of accounts dropped due to each step in the sample restrictions, together with the number of login events and sell events (account × stock × days) dropped at each step. From the starting sample of approximately 92,000 accounts, the largest drop of accounts is due to dropping approximately 29,000 inactive accounts (31.6% of accounts). After applying all six sample restrictions the resulting baseline sample retains 58% of accounts from the unrestricted sample. Our sample restrictions tend to drop accounts with below-average logins and sales (due to the largest drop being the drop of inactive accounts), hence the baseline sample retains 63% of login-days and 69% of sell-days.

As a second step, we restrict to a sample for analysis. Two motivations drive our sample selection. First, responses to changes in left-digits are only detectable in a sample of observations for an investor in which the left-digit changes. A key element in our analysis therefore is to draw a "price increasing sample" and a "price decreasing sample", which we define below. Moreover, we show that the response of changes in left-digit is very different depending upon when the stock is increasing in value or decreasing in value over time, in particular, selling activity occurs when prices cross left-digits from below and from above.

Second, responses to changes in left-digits are contingent upon the investor observing the change in left-digit. For example, a stock that changes left-digit over a holding period

 $<sup>^{2}</sup>$  In cases where the account satisfies this sample restriction in other years, we keep those years of observations in the data set.

in which the investor does not make a login to the account is much less likely to be noticed compared with a change in left-digit which occurs between login days within a holding period. We therefore apply sample restrictions in order to obtain a series of observations in which the price crosses the left-digit between login-days.

We define the price increasing sample and the price decreasing sample as follows. First, using the example of the price increasing sample, we identify the first day in each calendar quarter on which an investor made a login to their account.<sup>3</sup> We then define the price increasing sample as the set of login days within the quarter for which the prices on subsequent login days were always above the price on the first day and the left-digit had changed within the quarter on at least one subsequent login-day. This sample therefore provides a series of login-days through the quarter in which the price of the stock had broached a left-digit change on at least one of the login-days.

We define the price decreasing sample using parallel sample restrictions applied to decreasing prices. Hence the price decreasing sample is defined as the set of login days within the quarter for which the prices on subsequent login days were always below the price on the first day and the left-digit had changed within the quarter on at least one subsequent login-day. Our samples are based on quarters and individual × login days during the quarter.

#### 1.2 Summary Statistics

Table 2 provides summary statistics for account holder characteristics and account characteristics in the baseline sample. Approximately 81% of account holders are male and the average age of an account holder is 55 years. Account holders have held their accounts with Barclays for, on average, approximately five years, with approximately 25% of account holders having held their account for over seven years. The average portfolio value is approximately £71,000 (median £19,000), with accounts containing of average 6 stocks (median 4 stocks). Investors in the baseline sample overwhelmingly hold positions in a few common stocks. Holdings of mutual funds account for only 8% of the average investor's portfolio. Investors in the sample login approximately once per five days, but trade much less frequently at a frequency of

<sup>&</sup>lt;sup>3</sup> We show later that results are unchanged when we modify the period that defines a sample to either a month, or a year, instead of a quarter.

approximately once every thirty days.

Table 3 describes the price data for the baseline sample, price increasing sample and price decreasing sample. The baseline sample provides approximately 84.6 million login-day observations (the bottom row of Table 1). Panel A summarises prices of all observations paired with login-days and sell-days in the first two rows, together with price of stocks sold in the third row. The mean price of a stock in the sample of login-days is approximately £87, with a median of £3.

Panels B and C summarise prices for stocks from observations in the price increasing sample and observations in the price decreasing sample. Note, there are four units of left-digit in the data, pennies, tens of pennies, pounds and tens of pounds (there are only a few cases of hundreds of pounds). So, the left-digit changes of interest are pence to tens of pence, tens of pence to pounds, and pounds to tens of pounds (plus a few cases of tens of pounds to hundreds of pounds). The most common price range for observations in both the price increasing sample and the price decreasing sample is the £1.1 to £10.1 range, which accounts for 54.8% of observations in the price increasing sample and 43.7% of observations in the price decreasing sample.  $^4$ 

## 2 Results

#### 2.1 Main Results

Our main result is shown in Figure 1, which shows the probability of selling a stock by the leftmost price digit of the stock. The left-side plots in panels A and B take the samples of observations from the price increasing sample and price decreasing sample and stack observations (investor  $\times$  stock  $\times$  login days) by the leftmost two digits of the observation, centred on the change in left digit between some integer X and the next Y = X + 1. For example, an investor holds a stock and, in the price increasing sample, sees prices on login-days of 18p, 22p and 26p. These observations stack onto the x-axis at X8, Y2 and Y6, centring on the change in left-digit from 1 to 2 with X = 1, Y = 2. Another investor holds a stock, in the price increasing sample,

<sup>&</sup>lt;sup>4</sup> Most stocks in the samples are prices in the range £1.10 to £10.10. A histogram of prices for all investor  $\times$  login days is shown in Figure A1.

sees prices on login days of 361p, 389p and 430p. These observations stack onto the x-axis at X6, X8 and Y3, centring on the change in left-digit from 3 to 4, with X=3,  $Y=4.^5$  One stock  $\times$  day can contribute to multiple observations, if more than one investor makes a login to their account on the day. By way of contrast, a stock  $\times$  day might contribute no observations if no investor makes a login to their account on the day.

The first clear pattern in the left-side plots is that the probability of sale jumps upwards when the stock crosses a left-digit from below in the price increasing sample, and the probability of sale jumps downwards when the stock crosses a left-digit from above in the price-decreasing sample. In the price increasing sample, the increase is from approximately 0.008 to 0.0014, a 75% increase, whereas in the price decreasing sample, the increase is from approximately 0.008 to 0.0012, a 50% increase.<sup>6</sup>

A second clear pattern in the left-side plot is that the probability of sale jumps upwards when the price crosses the left-digit, and remains elevated for the next adjoining units, in the price increasing sample Y1, Y2, Y3, ..., in the price decreasing sample X9, X8, X7, ... One explanation for this pattern is that the investor's response to a change in left-digit does not necessarily occur when the price crosses Y0, as the investor may not observe the stock at price Y0 if they do not make a login at that point in time. Instead, the investor may only observe the change in left-digit when they log in and by the point the stock reaches Y1, Y2, Y3, ... in the price increasing sample for example. by the point the stock reaches, e.g. in the price increasing sample Y1, Y2, Y3, ...

The right-side plots in each panel illustrate the probability of sale across the full range of leftmost two digits of prices in the sample, unstacking the data from the left-side plot by leftmost two digits. In the example above, the series of prices 18p, 22p and 26p enter into the x-axis bins for 18, 20 and 26; while the series of prices 361p, 389p and 430p enter into the ex-axis bins for 36, 38 and 43. In both panels, each bin in which the right-digit of the leftmost two digits ends 9 is colored blue, with each bin in which the right-digit of the leftmost two digits ends 0 is colored red.

<sup>&</sup>lt;sup>5</sup> Stocks that cross multiple left-digits in the quarter contribute observations centred around each left-digit change.

<sup>&</sup>lt;sup>6</sup> The probability of sale is higher in the price increasing sample compared with the price decreasing sample, indicating that investors are more likely to sell stocks when prices increase compared with when prices decrease.

The clear pattern in the right-side plots is that the probability of sale jumps when the price crosses a leftmost digit from below in Panel A, and from above in Panel B. This is seen across the broad range of leftmost two digits of prices. The plot also makes clear the "ladder" pattern whereby the elevated probability of sale at each crossing of the leftmost digit steps down as the price moves further upwards (in Panel A) or downwards (in Panel B), with the probability jumping again at the next change in leftmost digit.

This pattern in the probability of sale when a stock price crosses a leftmost digit is *not* seen when one pools all observations. Pooling all observations stacks observations from price increasing series, price decreasing series and observations left out from quarters where prices did not show an stable increasing/decreasing pattern or remained largely unchanged. Figure A2 reproduces the figures from Figure 1 in the sample of all observations [JUST TO CHECK WE ARE STACKING UP PANELS A AND B OF THE MAIN FIGURE? [All obs, more than 80 million, including panels A and B and remaining]. The left-side plot shows no discontinuity in the probability of sale when price crosses the leftmost digit. In the figure, the probability of sale is slightly higher at lower prices, but the difference in probability across the x-axis very small (in the range 0.0079 to 0.0086 0.0085), compared with the jump in probability at the crossing of the left-digit of 0.006 in the price increasing sample and 0.004 in the price decreasing sample. The right-side plot also shows no difference in the probability of sale between the red bars and blue bars.

We estimate the size of the left-digit effect in Table 4 and Table 5, which show regression estimates for the price increasing sample and price decreasing samples respectively. The dependent variable in each regression is a dummy variable for whether the observation is a sale (either partial or total sale). The specification in Column 1 includes only a dummy for whether the stock price is *Y*0 or above. The coefficient on the dummy variable is 0.0044 in the price increasing sample. This implies an increase in the probability of sale when the stock price crosses the left-digit from below, evaluated against the constant term, of 52%. The coefficient on the dummy variable is -0.0024 in the price decreasing sample. This implies an increase in the probability of sale when the stock price crosses the left-digit from above, evaluated against the constant term, of 24%.

Additional columns in both tables add further controls to the econometric specification. The specification in Column 2 of each table adds slope terms for the range Y0-Y5 and X6-X9. Subsequent Columns add a series of fixed effects: day fixed effects, industry fixed effects (using industry classifications based on Datastream Industry Classification Benchmark (ICB)), account fixed effects and finally stock fixed effects. The specification in Column 5 therefore exploits within-investor, within-stock variation in the probability of sale, conditioning on day differences in the likelihood of sale. In this richest specification, the coefficients on the dummy variable of 0.0061 and -0.0040 imply an increase in the probability of sale when the stock price crosses the left-digit from below of 71% and an increase in the probability of sale when the stock prices crosses the left-digit from above of 40%.

The main results in Figure 1 pool over leftmost digits in pence, pounds and tens of pounds. 8. In further analysis, we reproduce the figures show in Figure 1 and regression estimates in Table 4 and Table 5 for separate samples in which stock prices are in pence, pounds and tens of pounds. Figure 2 and Figure 3 each show three panels with the sample split into observations with prices in the range £0.11 to £1.01 in Panel A, £1.01 to £10.1 in Panel B, and £11 to £101 in Panel C of each figure. The number of observations is uneven distributed across these subsample, with approximately 50% of observations from the main samples in Panel B of each figure and only 7% of observations from the main sample in Panel C of each figure. The jump in probability of sale when the price crosses the leftmost digit from below, or when the price crosses the leftmost digit from above, is seen in each sub-sample. 9 Table A1 and Table A2 report regressions for the subsamples by pennies, pounds and tens of pounds. Estimates from these models show very similar results to the main regression estimates.

 $<sup>^{7}</sup>$  These percentage increases in the probability of sale are calculated from the coefficient on the constant term in Column 1 of each table.

<sup>&</sup>lt;sup>8</sup> There are very few observations of stock prices in which the leftmost digit is in hundreds of pounds.

<sup>&</sup>lt;sup>9</sup> The jump in probability of sale when the leftmost digit crosses zero is less clear in Panel C of each plot, which contains the highest priced stocks in each sample, and contributed the smallest percentage of observations in each main sample.

#### 2.2 Robustness Tests

#### 2.2.1 Limit Orders

Why do we observe the jump in probability of sale when the price crosses a left-digit? One potential explanation for the pattern seen in the price increasing sample is that investors set limit orders with strike prices at round numbers (e.g., base-10 or base-100 numbers). Two observations of the patterns seen in Figure 1 suggest this is unlikely to explain the pattern we observe in full. First, while the use of limit orders might contribute to the pattern observed in Panel A, it would not explain the pattern observed in Panel B. Second, in liquid markets we would expect limit order to be executed at the strike price which, if a round number, would cause a spike in the probability of sale only at Y0 in the illustrations, not an elevated probability of sale at Y1, Y2, Y3, .... There are, however, two counter-arguments to this. First, if individuals place limit order outside of trading hours, the price may have risen further above Y0 by the time the brokerage executes the order (overnight orders form a queue on the broker's order book, and orders appearing later in the queue may execute at a price further away from the strike price). Second, if the stock is illiquid the brokerage may only be able to execute the order once the price has risen further above Y0 (again, due to queueing).

As a first set of robustness tests, we therefore exclude observations of sales which have a higher probability of being placed as a limit order. For example, the share of limit orders in the sample of orders placed out-of-hours is likely to be higher than the share of limit orders in the sample of orders placed during market hours. We exclude two sets of observations. First, observations of sales where the order is completed by the broker outside UK market hours (8am to 4.30pm). Second, observations of sales where the investor may a login to their account on the previous day , but no sale was executed on the previous day (but may have placed a sell order using a limit order) and may have placed a sell order using a limit order 10. Panels A and B of Figure 4 show the impact of implementing these exclusions. The pattern seen in the main analysis, of a jump in the probability of sale when the price crosses the leftmost digit from below in the price increasing sample, is seen in both panels.

Second, we restrict the sample to only the most liquid stocks, focusing on stocks in the

 $<sup>^{10}</sup>$  Likewise, we also exclude observations of sales executed on Mondays when the investor  $\log$  in during the weekend

FTSE100 only. In this sample, the likelihood of round-number limit orders executing at their strike price is higher compared with less liquid stocks in the FTSE250 or FTSE All-Share indices. If limit orders fulfilling at round number strike prices are responsible for the jump in probability of sale when the leftmost digit crosses zero from below, we should therefore see a sharp "spike" in sales at Y0 in the sample. Panel C, which restricts to FTSE100 stocks only, shows an elevated probability of sales at prices Y1, Y2, Y3, ... in the same way as seen in the main results. This again suggests that the pattern we see in the main results is not explained by investors using limit orders.

Third, we follow an approach to identifying limit order trades suggested by Linnainmaa (2010). Linnaimaa (2010) explores the role of limit orders as a potential explanation for the existence of the disposition effect, arguing that the disposition effect (among other features of individual investor trading behavior) can be explained by investors' use of limit orders. The paper proposes a method for identifying trades more likely to have been executed via limit orders.

The approach proceeds as follows. We first take all observations of buys and sell events for each investor in the price increasing sample. We then regress a buy-versus-sell indicator (a dependent variable that takes the value of one when an investor sells a stock and the value of zero when an investor purchases a stock) against the daily return of an stock, for each investor. Following Linnaimaa, we included investors with at least 10 trades. According to Linnaimaa's method, the same-day return coefficient is significantly positive for limit-order trades, but significantly negative for market-order trades (because individuals who are net buyers when the stock price falls, and net sellers when the stock price rises, are likely to be limit-order traders; while individuals who submit market orders often trade in the direction of the same-day return, and hence against limit order traders). Using this method, we are able to exclude from the sample investors for whom a positive coefficient is estimated from the analysis (839 investors and 6,144 sell events, 11% of the total number of account × stock × sell days [This is the reduction from the increasing sample, I think before I stated the reduction from all the observations in the data, which is not accurate. We do not have that many potential cases of limit orders and still an 11% of reduction of sales is a good reduction I think. And

these people excluded are usually frequent traders. Perhaps we can just quote the reduction in observations rather than the reduction of investors??]). <sup>11</sup>

We show the effect of excluding limit order investors on our main results in Panel D of Figure 4. The pattern seen is the main sample is unchanged in this further restricted sample, with a clear increase in the probability of sale when the price increases through a change in the leftmost digit. We show regression estimates using each of the samples in Figure 4 in Table A4 and Table A4. The regression results show the same patterns as those using the main sample.

#### 2.3 Simulation

The main results are shown using samples of stocks which increase, or decrease, in price, crossing a left-digit within the period of observations (a calendar quarter). This sample selection therefore contains a large proportion of observations with prices at or just above *Y*0 in the price increasing sample, or at or just below *Y*0 in the price decreasing sample, as this is the minimum criteria for inclusion in the price increasing sample, or price decreasing sample, respectively. Hence, a large share of the total number of observations of sales in each sample is clustered around *Y*0.

This sample selection should not mechanistically create a higher *probability* of selling a stock at Y0 compared with other leftmost two digit prices in the range X6-Y5 used in the analysis. However, to test for this we conduct a simulation analysis in which we assign sales randomly to investor  $\times$  stock  $\times$  days in each sample based on the average probability to sell observed in the data. [I assigned the probability to sell to all the observations based on the average probability to sell in the whole data. But of course we only use the samples here]. If our main results are due to sample selection, we should see the jumps in probability of sale when sells are randomly assigned across observations. Figure A3 shows that with randomly allocated sales we see no evidence of discontinuity in the probability of sale when the leftmost digit changes. This suggests that our main result does not arise mechanistically due to the sample selection.

<sup>&</sup>lt;sup>11</sup> Using trading records of all investors in Finland for the period 1995-2002, Linnaimaa shows that the fraction of actual limit-order trades in the sample of investors with positive coefficient, following this approach, is 65%

## 2.4 Sensitivity Tests

#### 2.4.1 Variation in Time Period

We test the sensitivity of our main estimates to the time period over which changes in the leftmost digit are calculated in the price increasing sample and the price decreasing sample. In our main analysis this time period is a calendar quarter, with observations restricted to the set of login days within the quarter for which the prices on subsequent days were always above the price on the first day and the left-digit had changed within the quarter at least once on a subsequent login-day.

We show the sensitivity of our results to shortening the time period for this sample restrict to either one month, or extending the sample restriction to one year in Figure A4 and as a year in Figure A5. The patterns seen in these figures are very similar to those seen in the main analysis. Table A5 reports for summary statistics for the stock prices in the monthly and yearly samples, with Table A6 and Table A7 showing regression estimates.

## 2.4.2 Sell-Day Sample

We also test the sensitivity of the main results to restricting the sample to sell-days instead of login-days. One could conduct the analysis of all days, login-days, sell-days or some other restriction to types of days. In our main analysis we use login-days as these are days on which the probability of an investor paying attention to the prices of stocks in their portfolio is higher, because they make a login to the account. We further restrict to sell-days as on sell-days we might be expect an even higher probability that the investor pays attention to the prices of stocks in the portfolio, given that they make a sale.

We show results when restricting the baseline sample to sell-days in Figure A6. The same pattern is seen as in the main analysis, of a jump in the probability of sale when the price crosses the leftmost digit from below in the price increasing sample, and a jump in the probability of sale when the prices crosses the leftmost digit from above in the price decreasing sample. Regression estimates are shown in Table A8 and Table A9. We also show results from this analysis in sub-samples by pennies, pounds and tens of pounds in Figure A7 and Figure A8, with regression analysis for these sub-samples shown in Table A10 and Table A11.

## 3 Who Exhibits Left-Digit Bias?

We use sample splits and test for differences in the left-digit effect by various investor characteristics

- Age: stronger among younger investors (Table 6).
- Gender: no differences (Table 7).
- Portfolio value: stronger among small portfolios (Table 8).
- Tenure: stronger among younger accounts (Table 9).
- Number of Stocks: stronger with fewer stocks (Table 10).

Figure 1: Leftmost Stock Price Digit and Probability of Sale, Quarterly Sample

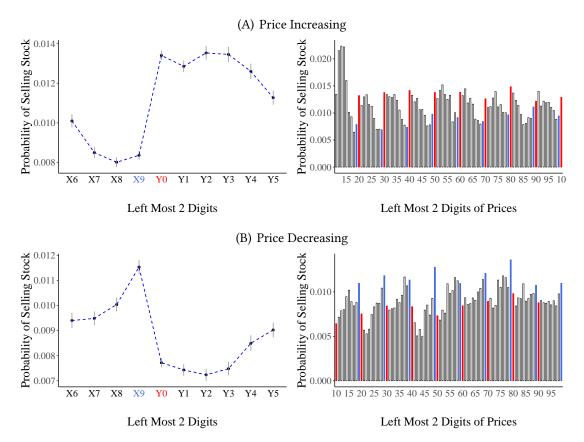
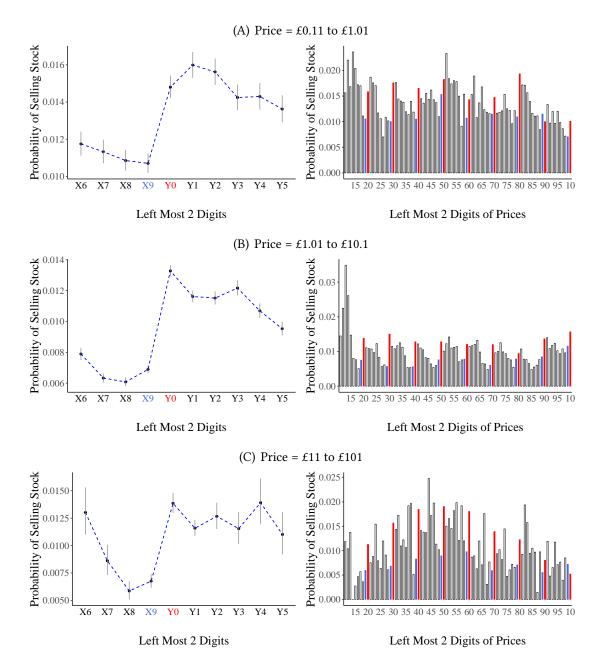
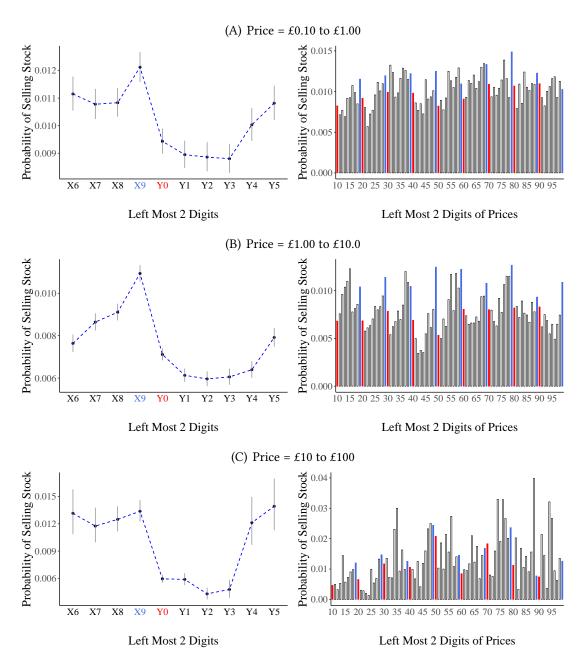


Figure 2: Leftmost Stock Price Digit and Probability of Sale Prices Increasing Sample by Price Range



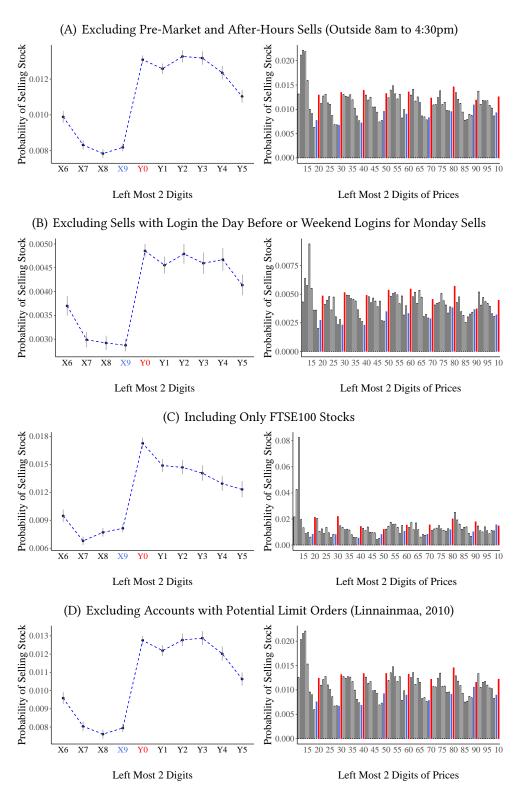
*Note:* £Y in the X-axes is equivalent to £X + 1 (e.g., £X9 could include £0.19, £1.9, £19, etc., while £Y0 could include £0.20, £2.0, £20, etc.). Panels A, B and C show equal size bins of 1p, 10p and £1, respectively. Panel A corresponds to 25% of the observations in the prices increasing sample; Panel B, to 55%; and Panel C, to 7%.

Figure 3: Leftmost Stock Price Digit and Probability of Sale Prices Decreasing Sample by Price Range



Note: £Y in the X-axes is equivalent to £X + 1 (e.g., £X9 could include £0.19, £1.9, £19, etc., while £Y0 could include £0.20, £2.0, £20, etc.). Panels A, B and C show equal size bins of 1p, 10p and £1, respectively. Panel A corresponds to 27% of the observations in the prices decreasing sample; Panel B, to 44%; and Panel C, to 7%.

Figure 4: Leftmost Stock Price Digit and Probability of Sale, Prices Increasing Sample Limit Order Robustness Tests



Note: £Y in the X-axes is equivalent to £X + 1 (e.g., £X9 could include £0.19, £1.9, £19, etc., while £Y0 could include £0.20, £2.0, £20, etc.). Panels display different sample restrictions to exclude sells corresponding to limit orders. Panel A drops sells executed when the market was closed. It also exclude potential discretionary trades (high frequency trades executed on the same stock, at the same time and at the same price that would likely correspond to sells arranged by Barclays discretionary service). Panel B excludes sells with a preceding login day. Panel C exclude non-liquid stocks, and Panel D excludes potential limit orders following Linnainmaa (2010) methodology. Panel A drops 2.2% of sells, Panel B drops 64.7% of sells, Panel C drops 75.9% of sells, and Panel D drops 11.4% of sells.

Table 1: Sample Selection

	Accounts	Logins	Sells
Unrestricted Sample	91817	135331214	993312
Drop due to:			
Inactive Accounts	28990	15951667	39075
Unmatched Prices	581	26014606	101667
At Least Two Stocks in Portfolio	5999	1444418	65638
Missing Demographic Data	2282	3980478	35724
Starting Position Days	40	726121	49899
Portfolio outliers	653	2631071	16339
Baseline sample	53272	84582853	684970

*Note:* The unrestricted sample contains 155,300 accounts. We use a 60% random sample of accounts. The table detail the steps in sample selection. Logins and Sells are at the account  $\times$  stock  $\times$  day level.

**Table 2: Baseline Sample Summary Statistics** 

	1		,			
	Mean	Min	p25	p50	p75	Max
A. Account Holder Characteristics						
Female	0.189					
Age (years)	54.848	17.000	47.000	57.000	67.000	87.000
Account Tenure (years)	5.291	0.060	3.063	4.049	6.959	16.975
B. Account Characteristics						
Portfolio Value (£10000)	7.106	0.000	0.654	1.911	5.347	265.820
Investment in Mutual Funds (£10000)	0.477	0.000	0.000	0.000	0.000	166.345
Investment in Mutual Funds (%)	7.862	0.000	0.000	0.000	0.000	12606.139
Number of Stocks	5.900	2.000	2.429	3.894	7.000	176.818
Login days (% all days)	18.749	0.076	4.261	11.528	28.603	100.000
Transaction days (% all market open days)	3.249	0.036	0.844	1.657	3.462	100.000
N Accounts	53272					

*Note:* This table presents summary statistics for the baseline sample of accounts. Age is measured at date of account opening. Account tenure is measured on the final day of the data period. Portfolio value is the value of all securities within the portfolio at market prices. Portfolio value, number of stocks and investment in mutual funds are measured as within-account averages of values at the first day of each calendar month in the data period. Login days is the percentage of days the account is open in the data period and the account holder made at least one login. Transaction days is the percentage of market open days the account is open in the data period and the account holder made at least one trade.

Table 3: Summary Stats, Quarterly Sample

Panei	(A):	Baseline	Sample

Price on Login Days £ Price on Sell Days £ Price of Stocks Sold £	N 84,582,853 6,390,539 684,970	Mean 7.392 6.626 7.069	St. Dev. 24.028 23.524 28.262	Min 0.000 0.000 0.000	Pctl(25) 1.144 0.835 0.844	Median 2.989 2.585 2.645	Pctl(75) 7.348 6.360 6.479	Max 15,051.630 3,589.000 2,057.994
		Panel (B)	: Price Inc	reasing S	Sample			
	N	Mean	St. Dev.	Min	Pctl(25)	Median	Pctl(75)	Max
All Stocks	4,869,714	6.035	20.802	0.000	0.738	2.961	6.074	3,589.000
Between £0.11 to £1.01	1,221,843	0.602	0.256	0.110	0.385	0.630	0.812	1.010
Between £1.1 to £10.1	2,670,842	4.864	2.305	1.100	2.947	4.519	6.550	10.100
Between £11 to £101	358,540	33.963	20.723	11.000	19.620	29.400	43.825	100.996
	Panel (C): Price Decreasing Sample							
	N	Mean	St. Dev.	Min	Pctl(25)	Median	Pctl(75)	Max
All Stocks	4,903,878	4.116	19.497	0.000	0.178	1.030	4.492	2,062.035
Between £0.10 to £1.0	1,346,143	0.513	0.270	0.100	0.276	0.486	0.755	1.000
Between £1 to £10	2,144,544	4.507	2.511	1.000	2.349	4.112	6.227	10.000
Between £10 to £100	340,298	24.961	17.954	10.000	10.870	20.510	29.940	99.990

Table 4: Probability of Sale and Left Digit, Price Increasing Sample

	$Probability\ of\ Sale_{ijt}=1$				
	(1)	(2)	(3)	(4)	(5)
Above Y0 = 1 (in Y0 - Y5)	0.0044***	0.0055***	0.0050***	0.0055***	0.0061***
Digits Y0 - Y5	(0.0001)	(0.0002) -0.0003***	(0.0002) -0.0004***	(0.0002) -0.0005***	(0.0002) -0.0008***
Digits X6 - X9		(0.0000) -0.0005***	(0.0000) -0.0003***	(0.0000) -0.0003***	(0.0000) -0.0002***
Constant	0.0086***	(0.0001) 0.0080***	(0.0001) 0.0077***	(0.0001)	(0.0001)
Constant	(0.0001)	(0.0001)	(0.0008)		
Day FE	NO	NO	YES	YES	YES
Industry FE	NO	NO	YES	YES	YES
Account FE	NO	NO	NO	YES	YES
Stock FE	NO	NO	NO	NO	YES
Observations	4,869,714	4,869,714	4,869,714	4,869,714	4,869,714
$\mathbb{R}^2$	0.0004	0.0004	0.0017	0.0679	0.0728

Table 5: Probability of Sale and Left Digit, Price Decreasing Sample

	$Probability\ of\ Sale_{ijt}=1$				
	(1)	(2)	(3)	(4)	(5)
Above Y0 = 1 (in Y0 - Y5)	-0.0024***	-0.0038***	-0.0042***	-0.0039***	-0.0040***
	(0.0001)	(0.0002)	(0.0002)	(0.0002)	(0.0002)
Digits Y0 - Y5		0.0002***	0.0003***	0.0004***	0.0004***
_		(0.0000)	(0.0000)	(0.0000)	(0.0000)
Digits X6 - X9		0.0007***	0.0008***	0.0006***	0.0006***
		(0.0001)	(0.0001)	(0.0001)	(0.0001)
Constant	0.0102***	0.0112***	0.0158***		
	(0.0002)	(0.0002)	(0.0012)		
Day FE	NO	NO	YES	YES	YES
Industry FE	NO	NO	YES	YES	YES
Account FE	NO	NO	NO	YES	YES
Stock FE	NO	NO	NO	NO	YES
Observations	4,903,878	4,903,878	4,903,878	4,903,878	4,903,878
$\mathbb{R}^2$	0.0002	0.0002	0.0005	0.0675	0.0717

Table 6: Probability of Sale and Left Digit, Splitting by Median Age

	Prices Increa	asing Sample	Prices Decre	asing Sample
	Below Median	Above Median	Below Median	Above Median
Above Y0 = 1 (in Y0 - Y5)	0.0075***	0.0047***	-0.0040***	-0.0040***
	(0.0003)	(0.0002)	(0.0002)	(0.0003)
Digits Y0 - Y5	-0.0009***	-0.0006***	0.0004***	0.0004***
_	(0.0001)	(0.0001)	(0.0001)	(0.0001)
Digits X6 - X9	-0.0003***	-0.0001	0.0007***	0.0004***
	(0.0001)	(0.0001)	(0.0001)	(0.0001)
Day FE	YES	YES	YES	YES
Industry FE	YES	YES	YES	YES
Account FE	YES	YES	YES	YES
Stock FE	YES	YES	YES	YES
Observations	2,580,896	2,288,818	2,654,464	2,249,414
$\mathbb{R}^2$	0.0866	0.0506	0.0869	0.0487

Table 7: Probability of Sale and Left Digit, Splitting by Gender

	Prices Increasing Sample		Prices Decre	easing Sample
	Female	Male	Female	Male
Above Y0 = 1 (in Y0 - Y5)	0.0064***	0.0061***	-0.0037***	-0.0040***
	(0.0004)	(0.0002)	(0.0004)	(0.0002)
Digits Y0 - Y5	-0.0006***	-0.0008***	0.0005***	0.0004***
	(0.0001)	(0.0000)	(0.0001)	(0.0000)
Digits X6 - X9	-0.0004***	-0.0001**	0.0005***	0.0006***
	(0.0002)	(0.0001)	(0.0002)	(0.0001)
Day FE	YES	YES	YES	YES
Industry FE	YES	YES	YES	YES
Account FE	YES	YES	YES	YES
Stock FE	YES	YES	YES	YES
Observations	837,223	4,032,491	787,316	4,116,562
$\mathbb{R}^2$	0.0700	0.0746	0.0703	0.0731

Table 8: Probability of Sale and Left Digit, Splitting by Portfolio Value

	Prices Increa	asing Sample	Prices Decre	asing Sample
	Below Median	Above Median	Below Median	Above Median
Above Y0 = 1 (in Y0 - Y5)	0.0090***	0.0035***	-0.0049***	-0.0030***
	(0.0003)	(0.0002)	(0.0003)	(0.0002)
Digits Y0 - Y5	-0.0011***	-0.0004***	0.0005***	0.0003***
-	(0.0001)	(0.0001)	(0.0001)	(0.0000)
Digits X6 - X9	-0.0003***	-0.0001	0.0008***	0.0002**
	(0.0001)	(0.0001)	(0.0001)	(0.0001)
Day FE	YES	YES	YES	YES
Industry FE	YES	YES	YES	YES
Account FE	YES	YES	YES	YES
Stock FE	YES	YES	YES	YES
Observations	2,390,422	2,479,292	2,496,374	2,407,504
$\mathbb{R}^2$	0.1030	0.0503	0.1063	0.0437

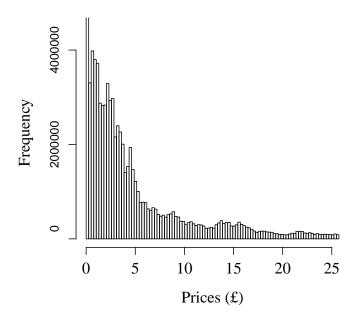
Table 9: Probability of Sale and Left Digit, Splitting by Account Tenure

	Prices Increa	asing Sample	Prices Decre	asing Sample
	Below Median	Above Median	Below Median	Above Median
Above Y0 = 1 (in Y0 - Y5)	0.0072***	0.0052***	-0.0046***	-0.0033***
	(0.0003)	(0.0002)	(0.0003)	(0.0003)
Digits Y0 - Y5	-0.0009***	-0.0006***	0.0005***	0.0003***
	(0.0001)	(0.0001)	(0.0001)	(0.0000)
Digits X6 - X9	-0.0002*	-0.0002**	0.0006***	0.0005***
	(0.0001)	(0.0001)	(0.0001)	(0.0001)
Day FE	YES	YES	YES	YES
Industry FE	YES	YES	YES	YES
Account FE	YES	YES	YES	YES
Stock FE	YES	YES	YES	YES
Observations	2,394,197	2,475,517	2,495,063	2,408,815
$\mathbb{R}^2$	0.0831	0.0607	0.0792	0.0645

Table 10: Probability of Sale and Left Digit, Splitting by Number of Stocks

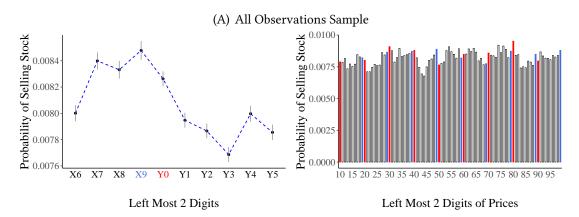
	Prices Increa	asing Sample	Prices Decre	asing Sample
	Below Median	Above Median	Below Median	Above Median
Above Y0 = 1 (in Y0 - Y5)	0.0088***	0.0031***	-0.0049***	-0.0030***
	(0.0003)	(0.0002)	(0.0003)	(0.0002)
Digits Y0 - Y5	-0.0011***	-0.0004***	0.0005***	0.0004***
	(0.0001)	(0.0001)	(0.0001)	(0.0000)
Digits X6 - X9	-0.0003***	-0.0001	0.0009***	0.0002**
	(0.0001)	(0.0001)	(0.0001)	(0.0001)
Day FE	YES	YES	YES	YES
Industry FE	YES	YES	YES	YES
Account FE	YES	YES	YES	YES
Stock FE	YES	YES	YES	YES
Observations	2,650,726	2,218,988	2,492,742	2,411,136
$\mathbb{R}^2$	0.0908	0.0377	0.0934	0.0329

Figure A1: Histogram of Stock Prices



*Note:* Figure shows the histogram of prices on login days. Outliers above the 95 percentile are excluded.

Figure A2: Leftmost Stock Price Digit and Probability of Sale, Quarterly Sample



Note: £Y in the X-axes is equivalent to £X + 1 (e.g., £X9 could include £0.19, £1.9, £1.9, £1.9, etc., while £Y0 could include £0.20, £2.0, £20, etc.). The figure exclude an atypical spike in sells on 2014-02-26 of stock US92343V1044 (99.99% of the total number of sales executed for that stock during the sample period).

Figure A3: Sample Selection and Simulation Exercise

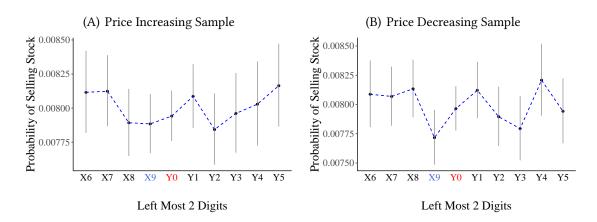


Figure A4: Leftmost Stock Price Digit and Probability of Sale, Monthly Sample

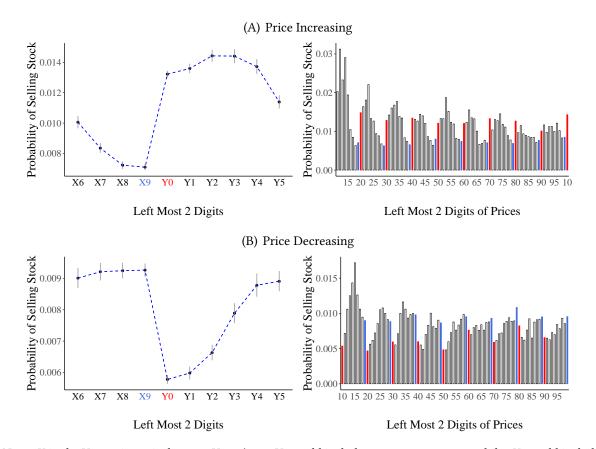


Figure A5: Leftmost Stock Price Digit and Probability of Sale, Annual Sample

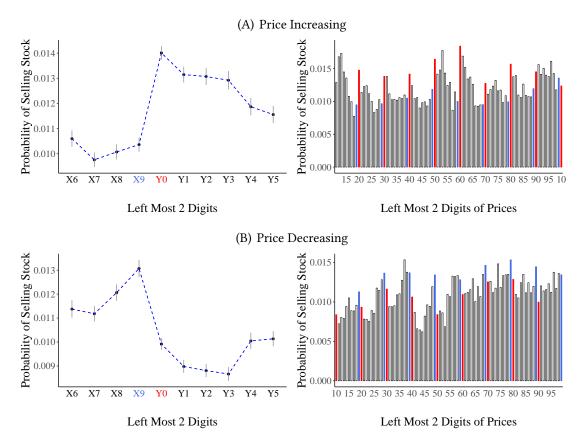


Figure A6: Leftmost Stock Price Digit and Probability of Sale, Sell Days

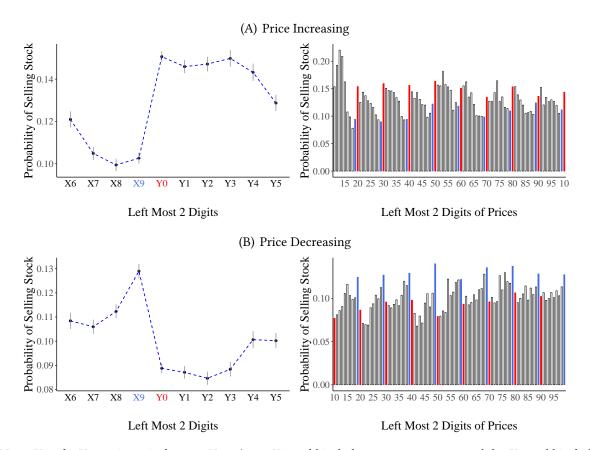
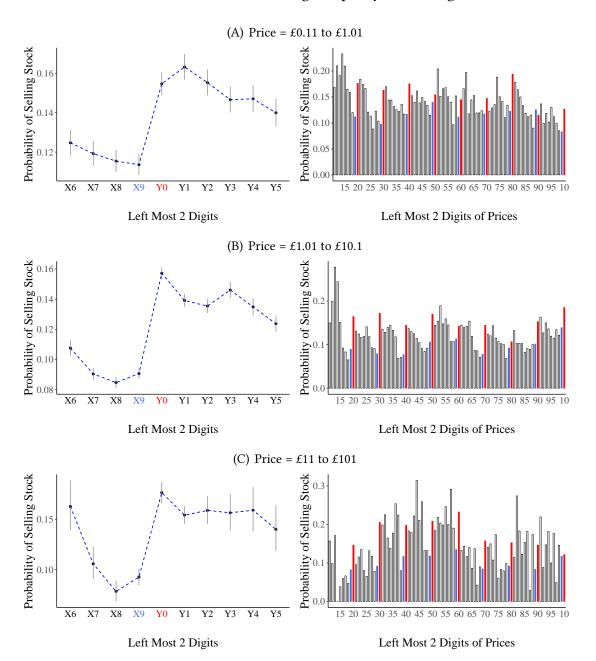
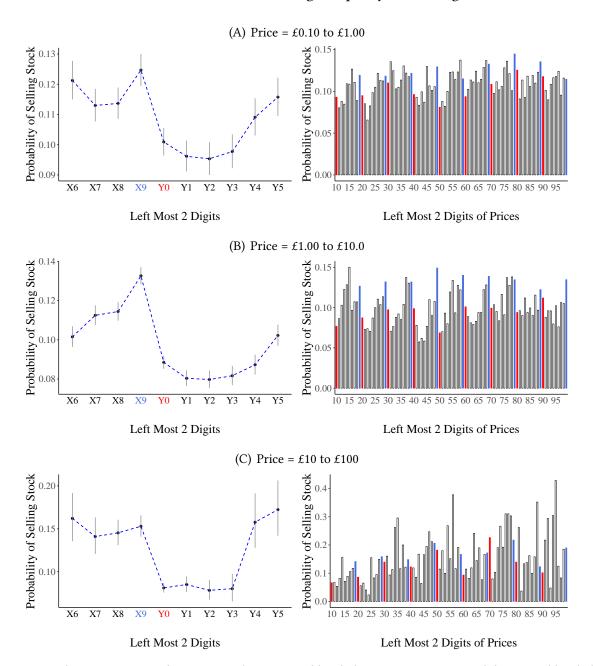


Figure A7: Leftmost Stock Price Digit and Probability of Sale, Sell Days Prices Increasing Sample by Price Range



*Note:* £Y in the X-axes is equivalent to £X + 1 (e.g., £X9 could include £0.19, £1.9, £19, etc., while £Y0 could include £0.20, £2.0, £20, etc.). Panels A, B and C show equal size bins of 1p, 10p and £1, respectively.

Figure A8: Leftmost Stock Price Digit and Probability of Sale, Sell Days Prices Decreasing Sample by Price Range



Note: £Y in the X-axes is equivalent to £X + 1 (e.g., £X9 could include £0.19, £1.9, £19, etc., while £Y0 could include £0.20, £2.0, £20, etc.). Panels A, B and C show equal size bins of 1p, 10p and £1, respectively.

Table A1: Price Increasing Subsamples with Equal Prices Bins

Panel	(A): Price	= f0.11	to £1.01

	$Probability\ of\ Sale_{ijt}=1$							
	(1)	(2)	(3)	(4)	(5)			
Above Y0 = 1 (in Y0 - Y5)	0.0037***	0.0049***	0.0046***	0.0048***	0.0047***			
,	(0.0002)	(0.0003)	(0.0003)	(0.0004)	(0.0004)			
Digits Y0 - Y5		-0.0003***	-0.0003***	-0.0004***	-0.0006***			
		(0.0001)	(0.0001)	(0.0001)	(0.0001)			
Digits X6 - X9		-0.0004***	-0.0002	-0.0003**	-0.0003**			
		(0.0001)	(0.0001)	(0.0001)	(0.0001)			
Constant	0.0111***	0.0106***	0.0221***					
	(0.0003)	(0.0003)	(0.0031)					
Day FE	NO	NO	YES	YES	YES			
Industry FE	NO	NO	YES	YES	YES			
Account FE	NO	NO	NO	YES	YES			
Stock FE	NO	NO	NO	NO	YES			
Observations	1,221,843	1,221,843	1,221,843	1,221,843	1,221,843			
$\mathbb{R}^2$	0.0003	0.0003	0.0013	0.1013	0.1080			

Panel (B): Price = £1.01 to £10.1

	$Probability\ of\ Sale_{ijt}=1$							
	(1)	(2)	(3)	(4)	(5)			
Above Y0 = 1 (in Y0 - Y5)	0.0050***	0.0065***	0.0063***	0.0064***	0.0067***			
Digits Y0 - Y5	(0.0001)	(0.0002) -0.0006***	(0.0002) -0.0007***	(0.0002) -0.0006***	(0.0002) -0.0008***			
Digits X6 - X9		(0.0001) -0.0002***	(0.0001) -0.0001**	(0.0001) -0.0002***	(0.0001) -0.0001			
Constant	0.0067***	(0.0001) 0.0065***	(0.0001) 0.0134***	(0.0001)	(0.0001)			
	(0.0001)	(0.0002)	(0.0026)					
Day FE	NO	NO	YES	YES	YES			
Industry FE	NO	NO	YES	YES	YES			
Account FE	NO	NO	NO	YES	YES			
Stock FE	NO	NO	NO	NO	YES			
Observations	2,670,842	2,670,842	2,670,842	2,670,842	2,670,842			
$\mathbb{R}^2$	0.0007	0.0007	0.0020	0.0738	0.0764			

Panel (C): Price = £11 to £101

		$Probability\ of\ Sale_{ijt}=1$						
	(1)	(2)	(3)	(4)	(5)			
Above Y0 = 1 (in Y0 - Y5)	0.0052***	0.0068***	0.0069***	0.0085***	0.0091***			
	(0.0004)	(0.0005)	(0.0005)	(0.0006)	(0.0006)			
Digits Y0 - Y5		-0.0003	-0.0004*	0.0001	-0.0000			
		(0.0002)	(0.0002)	(0.0002)	(0.0002)			
Digits X6 - X9		-0.0015***	-0.0018***	-0.0015***	-0.0015***			
		(0.0003)	(0.0003)	(0.0003)	(0.0003)			
Constant	0.0073***	0.0061***	-0.0024***					
	(0.0003)	(0.0003)	(0.0007)					
Day FE	NO	NO	YES	YES	YES			
Industry FE	NO	NO	YES	YES	YES			
Account FE	NO	NO	NO	YES	YES			
Stock FE	NO	NO_	NO	NO	YES			
Observations	358,540	358,370	358,540	358,540	358,540			
$\mathbb{R}^2$	0.0006	0.0007	0.0030	0.1395	0.1444			

Table A2: Price Decreasing Subsamples with Equal Prices Bins

	Panel (A):	Price = £0.10	to £1.00					
	$Probability\ of\ Sale_{ijt}=1$							
	(1)	(2)	(3)	(4)	(5)			
Above Y0 = 1 (in Y0 - Y5)	0.0037***	0.0049***	0.0046***	0.0048***	0.0047***			
· ·	(0.0002)	(0.0003)	(0.0003)	(0.0004)	(0.0004)			
Digits Y0 - Y5	, ,	-0.0003***	-0.0003***	-0.0004***	-0.0006***			
_		(0.0001)	(0.0001)	(0.0001)	(0.0001)			
Digits X6 - X9		-0.0004***	-0.0002	-0.0003**	-0.0003**			
_		(0.0001)	(0.0001)	(0.0001)	(0.0001)			
Constant	0.0111***	0.0106***	0.0221***					
	(0.0003)	(0.0003)	(0.0031)					
Day FE	NO	NO	YES	YES	YES			
Industry FE	NO	NO	YES	YES	YES			
Account FE	NO	NO	NO	YES	YES			
Stock FE	NO	NO	NO	NO	YES			
Observations	1,221,843	1,221,843	1,221,843	1,221,843	1,221,843			
$\mathbb{R}^2$	0.0003	0.0003	0.0013	0.1013	0.1080			
	Panel (B):	Price = £1.00	to £10.0					
	(-)		bility of Sale	$e_{iit} = 1$				
	(1)	(2)	(3)	(4)	(5)			
Above Y0 = 1 (in Y0 - Y5)	-0.0027***	-0.0042***	-0.0045***	-0.0045***	-0.0042***			

	$Probability\ of\ Sale_{ijt}=1$						
	(1)	(2)	(3)	(4)	(5)		
Above Y0 = 1 (in Y0 - Y5)	-0.0027***	-0.0042***	-0.0045***	-0.0045***	-0.0042***		
	(0.0002)	(0.0002)	(0.0002)	(0.0002)	(0.0002)		
Digits Y0 - Y5		0.0001	0.0001*	0.0004***	0.0004***		
		(0.0000)	(0.0000)	(0.0001)	(0.0001)		
Digits X6 - X9		0.0011***	0.0011***	0.0007***	0.0007***		
		(0.0001)	(0.0001)	(0.0001)	(0.0001)		
Constant	0.0093***	0.0107***	0.0207***				
	(0.0002)	(0.0002)	(0.0080)				
Day FE	NO	NO	YES	YES	YES		
Industry FE	NO	NO	YES	YES	YES		
Account FE	NO	NO	NO	YES	YES		
Stock FE	NO	NO	NO	NO	YES		
Observations	2,144,544	2,144,544	2,144,544	2,144,544	2,144,544		
$\mathbb{R}^2$	0.0002	0.0003	0.0008	0.0797	0.0838		
		2.2000	2.2000	//	2.3000		

Panel (C): Price = £10 to £100

	$Probability \ of \ Sale_{ijt} = 1$						
	(1)	(2)	(3)	(4)	(5)		
Above Y0 = 1 (in Y0 - Y5)	-0.0069***	-0.0077***	-0.0069***	-0.0058***	-0.0046***		
	(0.0005)	(0.0006)	(0.0006)	(0.0007)	(0.0008)		
Digits Y0 - Y5		0.0006***	0.0007***	0.0006***	0.0003		
-		(0.0002)	(0.0002)	(0.0002)	(0.0002)		
Digits X6 - X9		0.0003	0.0007*	0.0001	0.0003		
		(0.0004)	(0.0004)	(0.0004)	(0.0004)		
Constant	0.0128***	0.0132***	0.0080***				
	(0.0006)	(0.0006)	(0.0017)				
Day FE	NO	NO	YES	YES	YES		
Industry FE	NO	NO	YES	YES	YES		
Account FE	NO	NO	NO	YES	YES		
Stock FE	NO	NO	NO	NO	YES		
Observations	340,298	340,298	340,298	340,298	340,298		
$\mathbb{R}^2$	0.0012	0.0012	0.0031	0.1497	0.1554		

Table A3: Price Increasing Sample Limit Order Robustness Tests

Panel (A): Excluding Pre-Market and After-Hours Sells (Outside 8am to 4:30pm)

	$Probability\ of\ Sale_{ijt}=1$							
	(1)	(2)	(3)	(4)	(5)			
Above Y0 = 1 (in Y0 - Y5)	0.0043***	0.0054***	0.0049***	0.0053***	0.0059***			
	(0.0001)	(0.0002)	(0.0002)	(0.0002)	(0.0002)			
Digits Y0 - Y5		-0.0003***	-0.0004***	-0.0005***	-0.0007***			
-		(0.0000)	(0.0000)	(0.0000)	(0.0000)			
Digits X6 - X9		-0.0005***	-0.0003***	-0.0002***	-0.0002***			
		(0.0001)	(0.0001)	(0.0001)	(0.0001)			
Constant	0.0084***	0.0078***	0.0075***					
	(0.0001)	(0.0001)	(0.0008)					
Day FE	NO	NO	YES	YES	YES			
Industry FE	NO	NO	YES	YES	YES			
Account FE	NO	NO	NO	YES	YES			
Stock FE	NO	NO	NO	NO	YES			
Observations	4,868,538	4,868,538	4,868,538	4,868,538	4,868,538			
$\mathbb{R}^2$	0.0004	0.0004	0.0018	0.0688	0.0736			

Panel (B): Excluding Sells with Login the Day Before or Weekend Logins for Monday Sells

	$Probability\ of\ Sale_{ijt}=1$							
	(1)	(2)	(3)	(4)	(5)			
Above Y0 = 1 (in Y0 - Y5)	0.0016***	0.0021***	0.0020***	0.0023***	0.0025***			
	(0.0001)	(0.0001)	(0.0001)	(0.0001)	(0.0001)			
Digits Y0 - Y5		-0.0001***	-0.0001***	-0.0002***	-0.0003***			
		(0.0000)	(0.0000)	(0.0000)	(0.0000)			
Digits X6 - X9		-0.0002***	-0.0002***	-0.0002***	-0.0002***			
		(0.0000)	(0.0000)	(0.0000)	(0.0000)			
Constant	0.0031***	0.0028***	0.0012***					
	(0.0001)	(0.0001)	(0.0003)					
Day FE	NO	NO	YES	YES	YES			
Industry FE	NO	NO	YES	YES	YES			
Account FE	NO	NO	NO	YES	YES			
Stock FE	NO	NO	NO	NO	YES			
Observations	4,834,411	4,834,411	4,834,411	4,834,411	4,834,411			
$\mathbb{R}^2$	0.0002	0.0002	0.0010	0.0616	0.0643			

Note: Panels A, B and C show equal size bins of 1p, 10p and £1, respectively. Panel A drops 0.018% of sells, Panel B drops 61% of sells, Panel C drops 76% of sells, and Panel D drops 11% of sells.

Table A4: Price Increasing Sample Limit Order Robustness Tests

Panel (C): Including Only FTSE100 Stocks

	$Probability \ of \ Sale_{ijt} = 1$						
	(1)	(2)	(3)	(4)	(5)		
Above Y0 = 1 (in Y0 - Y5)	0.0070***	0.0090***	0.0087***	0.0096***	0.0098***		
,	(0.0002)	(0.0003)	(0.0003)	(0.0004)	(0.0004)		
Digits Y0 - Y5	,	-0.0010***	-0.0009***	-0.0009***	-0.0009***		
O		(0.0001)	(0.0001)	(0.0001)	(0.0001)		
Digits X6 - X9		-0.0001	-0.0001	-0.0005***	-0.0005***		
O		(0.0001)	(0.0001)	(0.0001)	(0.0001)		
Constant	0.0079***	0.0077***	0.0259***	,	,		
	(0.0002)	(0.0002)	(0.0013)				
Day FE	NO	NO	YES	YES	YES		
Industry FE	NO	NO	YES	YES	YES		
Account FE	NO	NO	NO	YES	YES		
Stock FE	NO	NO	NO	NO	YES		
Observations	1,126,143	1,126,143	1,126,143	1,126,143	1,126,143		
$\mathbb{R}^2$	0.0010	0.0012	0.0025	0.1024	0.1031		
Panel (D): Excluding	Accounts with Potential Limit Orders (Linnainmaa, 2010)						
				`	., 2010)		
	(1)	Proba	bility of Sale	$e_{ijt} = 1$	•		
	(1)	Proba (2)	bility of Sale	$e_{ijt} = 1 \tag{4}$	(5)		
Above Y0 = 1 (in Y0 - Y5)	0.0042***	Proba (2) 0.0052***	(3) 0.0048***	$e_{ijt} = 1$ (4) $0.0053^{***}$	(5) 0.0059***		
, ,		Proba (2) 0.0052*** (0.0002)	(3) 0.0048*** (0.0002)	$e_{ijt} = 1$ (4) $0.0053^{***}$ (0.0002)	(5) 0.0059*** (0.0002)		
, ,	0.0042***	Proba (2)  0.0052*** (0.0002) -0.0003***	0.0048*** (0.0002) 0.0004***	$e_{ijt} = 1$ (4) $0.0053^{***}$ (0.0002) $-0.0005^{***}$	(5) 0.0059*** (0.0002) -0.0007***		
Digits Y0 - Y5	0.0042***	Proba (2)  0.0052*** (0.0002) -0.0003*** (0.0000)	0.0048*** (0.0002) -0.0004*** (0.0000)	$e_{ijt} = 1$ $(4)$ $0.0053^{***}$ $(0.0002)$ $-0.0005^{***}$ $(0.0000)$	(5) 0.0059*** (0.0002) -0.0007*** (0.0000)		
Digits Y0 - Y5	0.0042***	Proba (2)  0.0052*** (0.0002) -0.0003*** (0.0000) -0.0004***	0.0048*** (0.0002) -0.0004*** (0.0000) -0.0003***	$ \begin{aligned} \varepsilon_{ijt} &= 1 \\ (4) \\ 0.0053^{***} \\ (0.0002) \\ -0.0005^{***} \\ (0.0000) \\ -0.0003^{***} \end{aligned} $	(5) 0.0059*** (0.0002) -0.0007*** (0.0000) -0.0002***		
Digits Y0 - Y5 Digits X6 - X9	0.0042*** (0.0001)	Proba (2)  0.0052*** (0.0002) -0.0003*** (0.0000) -0.0004*** (0.0001)	0.0048*** (0.0002) -0.0004*** (0.0000) -0.0003*** (0.0001)	$e_{ijt} = 1$ $(4)$ $0.0053^{***}$ $(0.0002)$ $-0.0005^{***}$ $(0.0000)$	(5) 0.0059*** (0.0002) -0.0007*** (0.0000)		
Digits Y0 - Y5 Digits X6 - X9	0.0042*** (0.0001)	Proba (2)  0.0052*** (0.0002) -0.0003*** (0.0000) -0.0004*** (0.0001) 0.0076***	0.0048*** (0.0002) -0.0004*** (0.0000) -0.0003*** (0.0001) 0.0069***	$ \begin{aligned} \varepsilon_{ijt} &= 1 \\ (4) \\ 0.0053^{***} \\ (0.0002) \\ -0.0005^{***} \\ (0.0000) \\ -0.0003^{***} \end{aligned} $	(5) 0.0059*** (0.0002) -0.0007*** (0.0000) -0.0002***		
Digits Y0 - Y5 Digits X6 - X9 Constant	0.0042*** (0.0001) 0.0082*** (0.0001)	Proba (2)  0.0052*** (0.0002) -0.0003*** (0.0000) -0.0004*** (0.0001) 0.0076*** (0.0001)	0.0048*** (0.0002) -0.0004*** (0.0000) -0.0003*** (0.0001) 0.0069*** (0.0008)	$e_{ijt} = 1 $ $(4)$ $0.0053^{***}$ $(0.0002)$ $-0.0005^{***}$ $(0.0000)$ $-0.0003^{***}$ $(0.0001)$	(5) 0.0059*** (0.0002) -0.0007*** (0.0000) -0.0002*** (0.0001)		
Digits Y0 - Y5 Digits X6 - X9 Constant Day FE	0.0042*** (0.0001) 0.0082*** (0.0001) NO	Proba (2)  0.0052*** (0.0002) -0.0003*** (0.0000) -0.0004*** (0.0001) 0.0076*** (0.0001) NO	0.0048*** (0.0002) -0.0004*** (0.0000) -0.0003*** (0.0001) 0.0069*** (0.0008) YES	$e_{ijt} = 1$ (4) $0.0053^{***}$ (0.0002) $-0.0005^{***}$ (0.0000) $-0.0003^{***}$ (0.0001)	(5)  0.0059*** (0.0002) -0.0007*** (0.0000) -0.0002*** (0.0001)		
Digits Y0 - Y5  Digits X6 - X9  Constant  Day FE  Industry FE	0.0042*** (0.0001) 0.0082*** (0.0001) NO	Proba (2)  0.0052*** (0.0002) -0.0003*** (0.0000) -0.0004*** (0.0001) 0.0076*** (0.0001) NO	0.0048*** (0.0002) -0.0004*** (0.0000) -0.0003*** (0.0001) 0.0069*** (0.0008) YES	$e_{ijt} = 1$ (4) $0.0053^{***}$ (0.0002) $-0.0005^{***}$ (0.0000) $-0.0003^{***}$ (0.0001)  YES YES	(5)  0.0059*** (0.0002) -0.0007*** (0.0000) -0.0002*** (0.0001)  YES YES		
Digits Y0 - Y5  Digits X6 - X9  Constant  Day FE  Industry FE  Account FE	0.0042*** (0.0001) 0.0082*** (0.0001) NO NO	Proba (2)  0.0052*** (0.0002) -0.0003*** (0.0000) -0.0004*** (0.0001) 0.0076*** (0.0001) NO NO	0.0048*** (0.0002) -0.0004*** (0.0000) -0.0003*** (0.0001) 0.0069*** (0.0008) YES YES NO	$e_{ijt} = 1$ (4) $0.0053^{***}$ (0.0002) $-0.0005^{***}$ (0.0000) $-0.0003^{***}$ (0.0001)  YES YES YES	(5) 0.0059*** (0.0002) -0.0007*** (0.0000) -0.0002*** (0.0001) YES YES YES		
Digits Y0 - Y5  Digits X6 - X9  Constant  Day FE  Industry FE  Account FE  Stock FE	0.0042*** (0.0001) 0.0082*** (0.0001) NO NO NO	Proba (2)  0.0052*** (0.0002) -0.0003*** (0.0000) -0.0004*** (0.0001) 0.0076*** (0.0001) NO NO NO NO	0.0048*** (0.0002) -0.0004*** (0.0000) -0.0003*** (0.0001) 0.0069*** (0.0008) YES YES NO NO	$e_{ijt} = 1$ (4) $0.0053^{***}$ (0.0002) $-0.0005^{***}$ (0.0000) $-0.0003^{***}$ (0.0001)  YES YES YES NO	(5) 0.0059*** (0.0002) -0.0007*** (0.0000) -0.0002*** (0.0001) YES YES YES YES YES		
Above Y0 = 1 (in Y0 - Y5)  Digits Y0 - Y5  Digits X6 - X9  Constant  Day FE  Industry FE  Account FE  Stock FE  Observations  R <sup>2</sup>	0.0042*** (0.0001) 0.0082*** (0.0001) NO NO	Proba (2)  0.0052*** (0.0002) -0.0003*** (0.0000) -0.0004*** (0.0001) 0.0076*** (0.0001) NO NO	0.0048*** (0.0002) -0.0004*** (0.0000) -0.0003*** (0.0001) 0.0069*** (0.0008) YES YES NO	$e_{ijt} = 1$ (4) $0.0053^{***}$ (0.0002) $-0.0005^{***}$ (0.0000) $-0.0003^{***}$ (0.0001)  YES YES YES	(5)  0.0059*** (0.0002) -0.0007*** (0.0000) -0.0002*** (0.0001)  YES YES YES YES		

Note: Panels A, B and C show equal size bins of 1p, 10p and £1, respectively. Panel A drops 0.018% of sells, Panel B drops 61% of sells, Panel C drops 76% of sells, and Panel D drops 11% of sells.

Table A5: Summary Stats for Annual and Monthly Samples

	N	Mean	St. Dev.	Min	Pctl(25)	Median	Pctl(75)	Max
Monthly Increasing Sample	4,320,119	5.377	25.084	0.000	0.565	2.628	6.008	3,589.000
Monthly Decreasing Sample	5,124,666	4.625	24.242	0.000	0.211	1.002	5.038	2,062.035
Annual Increasing Sample	4,561,585	7.739	22.449	0.000	1.070	3.584	7.204	3,589.500
Annual Decreasing Sample	4,184,114	3.932	20.374	0.000	0.160	1.090	4.240	2,062.035

Table A6: Price Increasing Samples, Monthly and Annual Samples

Panel (A): Monthly Sample

	$Probability \ of \ Sale_{ijt} = 1$							
	(1)	(2)	(3)	(4)	(5)			
Above Y0 = 1 (in Y0 - Y5)	0.0057***	0.0067***	0.0062***	0.0065***	0.0070***			
	(0.0001)	(0.0002)	(0.0002)	(0.0002)	(0.0002)			
Digits Y0 - Y5		-0.0001*	-0.0002***	-0.0005***	-0.0007***			
		(0.0000)	(0.0000)	(0.0000)	(0.0000)			
Digits X6 - X9		-0.0009***	-0.0005***	-0.0001	-0.0001			
		(0.0001)	(0.0001)	(0.0001)	(0.0001)			
Constant	0.0078***	0.0069***	0.0102***					
	(0.0001)	(0.0001)	(0.0012)					
Day FE	NO	NO	YES	YES	YES			
Industry FE	NO	NO	YES	YES	YES			
Account FE	NO	NO	NO	YES	YES			
Stock FE	NO	NO	NO	NO	YES			
Observations	4,320,119	4,320,119	4,320,119	4,320,119	4,320,119			
$R^2$	0.0007	0.0007	0.0018	0.0632	0.0681			

Panel	(B	): Annua	l Sample
-------	----	----------	----------

	$Probability\ of\ Sale_{ijt}=1$						
	(1)	(2)	(3)	(4)	(5)		
Above Y0 = 1 (in Y0 - Y5)	0.0028***	0.0038***	0.0035***	0.0042***	0.0048***		
	(0.0001)	(0.0002)	(0.0002)	(0.0002)	(0.0002)		
Digits Y0 - Y5		-0.0005***	-0.0005***	-0.0005***	-0.0007***		
		(0.0000)	(0.0000)	(0.0000)	(0.0000)		
Digits X6 - X9		-0.0000	0.0000	-0.0002**	-0.0001		
-		(0.0001)	(0.0001)	(0.0001)	(0.0001)		
Constant	0.0102***	0.0102***	0.0089***				
	(0.0002)	(0.0002)	(0.0009)				
Day FE	NO	NO	YES	YES	YES		
Industry FE	NO	NO	YES	YES	YES		
Account FE	NO	NO	NO	YES	YES		
Stock FE	NO	NO	NO	NO	YES		
Observations	4,561,585	4,561,585	4,561,585	4,561,585	4,561,585		
$\mathbb{R}^2$	0.0002	0.0002	0.0025	0.0763	0.0812		

Table A7: Price Decreasing Samples, Monthly and Annual Samples

Panel (A): Monthly Sample

	$Probability\ of\ Sale_{ijt} = 1$						
	(1)	(2)	(3)	(4)	(5)		
Above Y0 = 1 (in Y0 - Y5)	-0.0024***	-0.0037***	-0.0040***	-0.0040***	-0.0042***		
	(0.0001)	(0.0001)	(0.0001)	(0.0001)	(0.0002)		
Digits Y0 - Y5	, ,	0.0007***	0.0007***	0.0006***	0.0006***		
		(0.0000)	(0.0000)	(0.0000)	(0.0000)		
Digits X6 - X9		0.0001	0.0002***	0.0002***	0.0003***		
		(0.0001)	(0.0001)	(0.0001)	(0.0001)		
Constant	0.0092***	0.0093***	0.0148***				
	(0.0002)	(0.0002)	(0.0010)				
Day FE	NO	NO	YES	YES	YES		
Industry FE	NO	NO	YES	YES	YES		
Account FE	NO	NO	NO	YES	YES		
Stock FE	NO	NO	NO	NO	YES		
Observations	5,124,666	5,124,666	5,124,666	5,124,666	5,124,666		
$\mathbb{R}^2$	0.0002	0.0003	0.0006	0.0590	0.0623		

Panel (	B)	١:	Annual	Sample
I and I	v	٠.	<i>i</i> minuai	Janipic

	$Probability\ of\ Sale_{ijt}=1$						
	(1)	(2)	(3)	(4)	(5)		
Above Y0 = 1 (in Y0 - Y5)	-0.0025*** (0.0001)	-0.0035*** (0.0002)	-0.0039*** (0.0002)	-0.0032*** (0.0002)	-0.0030***		
Digits Y0 - Y5	(0.0001)	0.0001	0.0000	0.0003***	(0.0002) 0.0003***		
Digits X6 - X9		(0.0000) 0.0006***	(0.0000) 0.0008***	(0.0000) $0.0004***$	(0.0000) $0.0004***$		
Constant	0.0119***	(0.0001) 0.0128***	(0.0001) 0.0168***	(0.0001)	(0.0001)		
Constant	(0.0002)	(0.0003)	(0.0012)				
Day FE	NO	NO	YES	YES	YES		
Industry FE	NO	NO	YES	YES	YES		
Account FE	NO	NO	NO	YES	YES		
Stock FE	NO	NO	NO	NO	YES		
Observations	4,184,114	4,184,114	4,184,114	4,184,114	4,184,114		
$\mathbb{R}^2$	0.0001	0.0002	0.0006	0.0796	0.0845		

Table A8: Probability of Sale and Left Digit, Price Increasing Sample, Sell Days

	$Probability \ of \ Sale_{ijt} = 1$					
	(1)	(2)	(3)	(4)	(5)	
Above Y0 = 1 (in Range Y0 to Y5)	0.0401*** (0.0019)	0.0523*** (0.0025)	0.0495*** (0.0025)	0.0424*** (0.0024)	0.0463*** (0.0024)	
Stock Digits Y0 to Y5	(0.001))	-0.0030***	-0.0036***	-0.0036***	-0.0057***	
		(0.0005)	(0.0005)	(0.0005)	(0.0005)	
Stock Digits X6 to X9		-0.0052***	-0.0038***	-0.0022***	-0.0015*	
		(0.0008)	(0.0008)	(0.0007)	(0.0008)	
Constant	0.1057***	0.0991***	0.0958***			
	(0.0032)	(0.0032)	(0.0092)			
Day FE	NO	NO	YES	YES	YES	
Industry FE	NO	NO	YES	YES	YES	
Account FE	NO	NO	NO	YES	YES	
Stock FE	NO	NO	NO	NO	YES	
Observations	418,946	418,946	418,946	418,946	418,946	
$\mathbb{R}^2$	0.0033	0.0036	0.0112	0.2464	0.2715	

*Note:* The unit of observation is an investor  $\times$  stock  $\times$  day. The samples is restricted to sell days. We include only quarters in which the stocks increased in price (regarding the first observation of the quarter) and change the left most digit at least once during the quarter. Only those stocks that have changed the left most digit are included. Regressions fit an intercept for the change in the left most digit at X0 and two slopes for the left (with values in the range -3 to 0, corresponding to X6 to X9) and right (with values in the range 0 to 5, corresponding to Y0 to Y5) values. The constant shows the probability to sell the stock at when the second digit is 9 (X9). The second digit over threshold dummy shows the jump in probability when the first digit changes and so the second digit becomes 0 (X0). SE are clustered by account.

Table A9: Probability of Sale and Left Digit, Price Decreasing Sample, Sell Days

	$Probability\ of\ Sale_{ijt}=1$					
	(1)	(2)	(3)	(4)	(5)	
Above Y0 = 1 (in Range Y0 to Y5)	-0.0240***	-0.0388***	-0.0414***	-0.0316***	-0.0313***	
Stock Digits Y0 to Y5	(0.0013)	(0.0021) $0.0024***$	(0.0020) 0.0025***	(0.0019) 0.0031***	(0.0019) 0.0037***	
Stock Digits X6 to X9		(0.0004) 0.0073***	(0.0004) 0.0078***	(0.0004) 0.0043***	(0.0004) 0.0035***	
C		(0.0009)	(0.0009)	(0.0008)	(0.0008)	
Constant	0.1147*** (0.0026)	0.1246*** (0.0031)	0.1554*** (0.0091)			
Day FE	NO	NO	YES	YES	YES	
Industry FE	NO	NO	YES	YES	YES	
Account FE	NO	NO	NO	YES	YES	
Stock FE	NO	NO	NO	NO	YES	
Observations	427,965	427,965	427,965	427,965	427,965	
$\mathbb{R}^2$	0.0015	0.0019	0.0035	0.2225	0.2450	

Note: The unit of observation is an investor  $\times$  stock  $\times$  day. The samples is restricted to sell days. We include only quarters in which the stocks have not increased in price (regarding the first observation of the quarter) and have not changed the left most digit at least once during the quarter. Regressions fit an intercept for the change in the left most digit at X0 and two slopes for the left (with values in the range -3 to 0, corresponding to X6 to X9) and right (with values in the range 0 to 5, corresponding to Y0 to Y5) values. The constant shows the probability to sell the stock at when the second digit is 9 (X9). The second digit over threshold dummy shows the jump in probability when the first digit changes and so the second digit becomes 0 (X0). SE are clustered by account.

Table A10: Price Increasing Subsamples with Equal Prices Bins, Sell Days

Panel	(A):	Price =	= £0.11	to	£1.01
-------	------	---------	---------	----	-------

	Probability of $Sale_{ijt} = 1$					
	(1)	(2)	(3)	(4)	(5)	
Above Y0 = 1 (in Range Y0 to Y5)	0.0341***	0.0471***	0.0438***	0.0298***	0.0273***	
	(0.0027)	(0.0038)	(0.0038)	(0.0037)	(0.0037)	
Stock Digits Y0 to Y5		-0.0035***	-0.0033***	-0.0029***	-0.0033***	
		(0.0009)	(0.0009)	(0.0009)	(0.0009)	
Stock Digits X6 to X9		-0.0036***	-0.0023	-0.0016	-0.0013	
		(0.0014)	(0.0014)	(0.0014)	(0.0015)	
Constant	0.1177***	0.1128***	0.2175***			
	(0.0045)	(0.0048)	(0.0242)			
Day FE	NO	NO	YES	YES	YES	
Industry FE	NO	NO	YES	YES	YES	
Account FE	NO	NO	NO	YES	YES	
Stock FE	NO	NO	NO	NO	YES	
Observations	117,610	117,610	117,610	117,610	117,610	
$\mathbb{R}^2$	0.0024	0.0026	0.0147	0.3492	0.3740	

Panel (B): Price = £1.01 to £10.1

	$Probability\ of\ Sale_{ijt}=1$					
	(1)	(2)	(3)	(4)	(5)	
Above Y0 = 1 (in Range Y0 to Y5)	0.0500***	0.0654***	0.0637***	0.0503***	0.0523***	
	(0.0026)	(0.0035)	(0.0035)	(0.0033)	(0.0033)	
Stock Digits Y0 to Y5	,	-0.0051***	-0.0058***	-0.0044***	-0.0062***	
		(0.0007)	(0.0007)	(0.0007)	(0.0007)	
Stock Digits X6 to X9		-0.0044***	-0.0036***	-0.0023**	-0.0008	
		(0.0010)	(0.0010)	(0.0010)	(0.0010)	
Constant	0.0919***	0.0866***	0.1168***	, ,	,	
	(0.0032)	(0.0031)	(0.0211)			
Day FE	NO	NO	YES	YES	YES	
Industry FE	NO	NO	YES	YES	YES	
Account FE	NO	NO	NO	YES	YES	
Stock FE	NO	NO	NO	NO	YES	
Observations	209,483	209,483	209,483	209,483	209,483	
$\mathbb{R}^2$	0.0057	0.0062	0.0141	0.3025	0.3176	

Panel (C): Price = £11 to £101

	$Probability\ of\ Sale_{ijt} = 1$					
	(1)	(2)	(3)	(4)	(5)	
Above $Y0 = 1$ (in Range $Y0$ to $Y5$ )	0.0642***	0.0849***	0.0811***	0.0582***	0.0619***	
	(0.0054)	(0.0068)	(0.0066)	(0.0077)	(0.0079)	
Stock Digits Y0 to Y5		-0.0052**	-0.0051**	0.0023	0.0014	
_		(0.0023)	(0.0023)	(0.0024)	(0.0026)	
Stock Digits X6 to X9		-0.0158***	-0.0172***	-0.0139***	-0.0117***	
		(0.0035)	(0.0035)	(0.0039)	(0.0040)	
Constant	0.0971***	0.0839***	-0.0061			
	(0.0041)	(0.0044)	(0.0133)			
Day FE	NO	NO	YES	YES	YES	
Industry FE	NO	NO	YES	YES	YES	
Account FE	NO	NO	NO	YES	YES	
Stock FE	NO	NO	NO	NO	YES	
Observations	27,460	$46_{7,460}$	27,460	27,460	27,460	
$\mathbb{R}^2$	0.0079	0.0090	0.0292	0.4648	0.4820	

Table A11: Price Decreasing Subsamples with Equal Prices Bins, Sell Days

		Proba	bility of Sale	$e_{ijt} = 1$	
	(1)	(2)	(3)	(4)	(5)
Above Y0 = 1 (in Range Y0 to Y5)	0.0341***	0.0471***	0.0438***	0.0298***	0.0273***
	(0.0027)	(0.0038)	(0.0038)	(0.0037)	(0.0037)
Stock Digits Y0 to Y5		-0.0035***	-0.0033***	-0.0029***	-0.0033***
		(0.0009)	(0.0009)	(0.0009)	(0.0009)
Stock Digits X6 to X9		-0.0036***	-0.0023	-0.0016	-0.0013
		(0.0014)	(0.0014)	(0.0014)	(0.0015)
Constant	0.1177***	0.1128***	0.2175***		
	(0.0045)	(0.0048)	(0.0242)		
Day FE	NO	NO	YES	YES	YES
Industry FE	NO	NO	YES	YES	YES
Account FE	NO	NO	NO	YES	YES
Stock FE	NO	NO	NO	NO	YES
Observations	117,610	117,610	117,610	117,610	117,610
$\mathbb{R}^2$	0.0024	0.0026	0.0147	0.3492	0.3740

Pa	anel (B): Pric	e = £1.00 to £	10.0					
		$Probability of Sale_{ijt} = 1$						
	(1)	(2)	(3)	(4)	(5)			
Above Y0 = 1 (in Range Y0 to Y5)	-0.0313***	-0.0472***	-0.0502***	-0.0388***	-0.0359***			
	(0.0019)	(0.0029)	(0.0029)	(0.0030)	(0.0030)			
Stock Digits Y0 to Y5		0.0018***	0.0019***	0.0037***	0.0031***			
		(0.0006)	(0.0006)	(0.0006)	(0.0007)			
Stock Digits X6 to X9		0.0099***	0.0100***	0.0046***	0.0048***			
		(0.0013)	(0.0012)	(0.0012)	(0.0012)			
Constant	0.1177***	0.1301***	0.2425***					
	(0.0030)	(0.0038)	(0.0795)					
Day FE	NO	NO	YES	YES	YES			
Industry FE	NO	NO	YES	YES	YES			
Account FE	NO	NO	NO	YES	YES			
Stock FE	NO	NO	NO	NO	YES			
Observations	166,494	166,494	166,494	166,494	166,494			
$\mathbb{R}^2$	0.0027	0.0033	0.0059	0.2875	0.3057			

	Panel (C): Price = £10 to £100  Prochability of Solo = 1							
	Probability of $Sale_{ijt} = 1$							
	(1)	(2)	(3)	(4)	(5)			
Above Y0 = 1 (in Range Y0 to Y5)	-0.0629***	-0.0735***	-0.0654***	-0.0427***	-0.0296***			
	(0.0059)	(0.0070)	(0.0070)	(0.0081)	(0.0098)			
Stock Digits Y0 to Y5		0.0108***	0.0099***	0.0053**	0.0017			
		(0.0024)	(0.0025)	(0.0027)	(0.0030)			
Stock Digits X6 to X9		0.0004	0.0030	-0.0047	-0.0029			
		(0.0049)	(0.0049)	(0.0052)	(0.0053)			
Constant	0.1497***	0.1501***	0.0929***					
	(0.0065)	(0.0068)	(0.0239)					
Day FE	NO	NO	YES	YES	YES			
Industry FE	NO	NO	YES	YES	YES			
Account FE	NO	NO	NO	YES	YES			
Stock FE	NO	NO	NO	NO	YES			
Observations	24,859	$47_{24,859}$	24,859	24,859	24,859			
$R^2$	0.0087	0.0102	0.0241	0.4411	0.4628			

## References

- Akepanidtaworn, K., R. Di Mascio, A. Imas, and L. Schmidt (2019). Selling fast and buying slow: Heuristics and trading performance of institutional investors. *Available at SSRN 3301277*.
- Poltrock, S. E. and D. R. Schwartz (1984). Comparative judgments of multidigit numbers. *Journal of Experimental Psychology: Learning, Memory, and Cognition 10*(1), 32.
- Shlain, A. S. (2018). More than a penny's worth: Left-digit bias and firm pricing. *manuscript*, *University of California, Berkeley*.