THE MESSAGE FILE SOLUTION



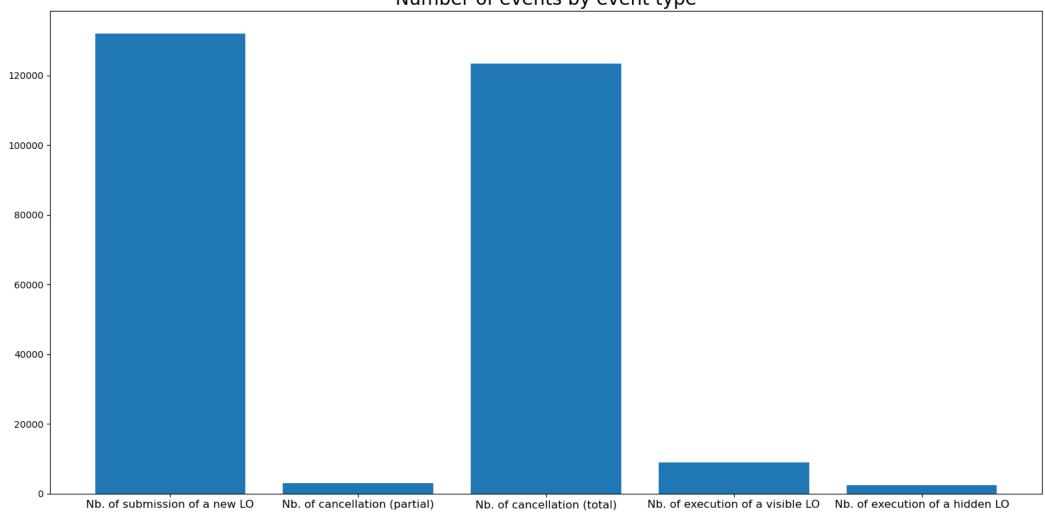
What I have found:

- Total number of event during the trading day 21/06/2012: 269 748
- Most active trading hour: 3-4 pm
- Second most active trading hour: 10-11 am
- Average nb. of event in one minute of trading: 683
- Min. nb. of order book event in one minute of trading: 0
- Max. nb. of order book event in one minute of trading: 2 127
- Imbalance 1: Visible limit order: -7%
- Imbalance 2: Hidden limit order: 10%
- Imbalance 3: Cancelled limit order: -3%
- Average time duration of cancelled trade: 5s
- Median time duration of cancelled trade: 0.7s

THE MESSAGE FILE

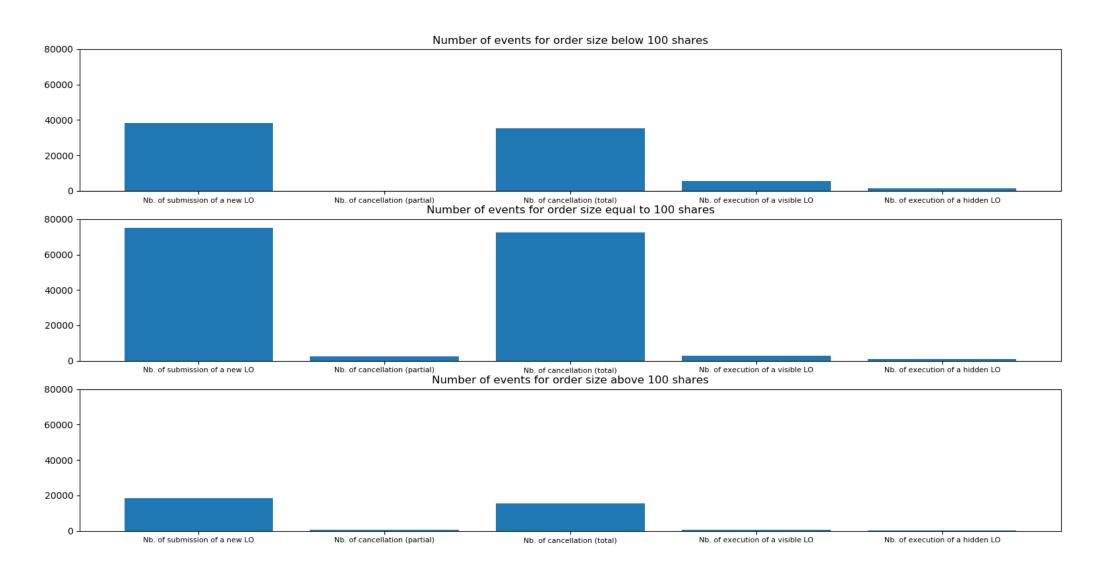






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THE MESSAGE FILE SOLUTION







Ask Price 1	Ask Size 1	Bid Price 1	Bid Size 1	Ask Price 2	Ask Size 2	Bid Price 2	Bid Size 2	
THICE I	JIZC I	THICE I	JIZC I	FIICC Z	JIZC Z	FIICC Z	JIZC Z	
÷	:	:	:	:	:	:	÷	÷
1186600	9484	118500	8800	118700	22700	118400	14930	
1186600	9384	118500	8800	118700	22700	118400	14930	
	:	:	:	:	:	:	:	÷

variable explanation.

Ask Price 1:Ask Size 1:Bid Price 1:Bid Size 1:Ask Price 2:	Level 1 ask price Level 1 ask volume Level 1 bid price Level 1 bid volume Level 2 ask price	(best ask price) (best ask volume) (best bid price) (best bid volume) (second best ask price)
Ask Price 2:Ask Size 2:	Level 2 ask price Level 2 ask volume	(second best ask price) (second best ask volume)

• ..

https://lobsterdata.com/info/ DataStructure.php

What is the link between the message file and the order book file?

« The k-th row in the 'message' file describes the limit order event causing the change in the limit order book from line k-1 to line k in the 'orderbook' file. »

PROGRAMMING EXERCISE 2 THE ORDER BOOK FILE



✓ Download the order book file (Courses). What is the total number of levels?

Step 1

- Show bid and ask level along the order book at the beginning of the trading day
- What is the cumulative market depth at the beginning of the trading day?

Step 2

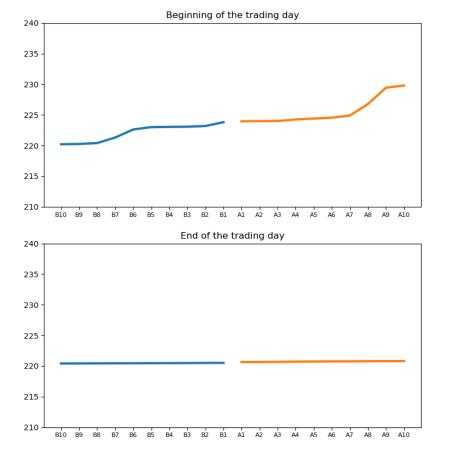
- Show bid and ask level along the order book at the middle of the trading day
- What is the cumulative market depth at the middle of the trading day?

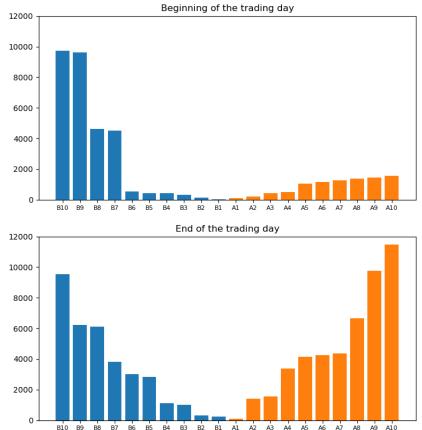
Step 3

- Show bid and ask level along the order book at the ending of the trading day
- What is the cumulative market depth at the ending of the trading day?

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PROGRAMMING EXERCISE 2 SOLUTION





PROGRAMMING EXERCISE 3 MESSAGE FILE + LOB FILE



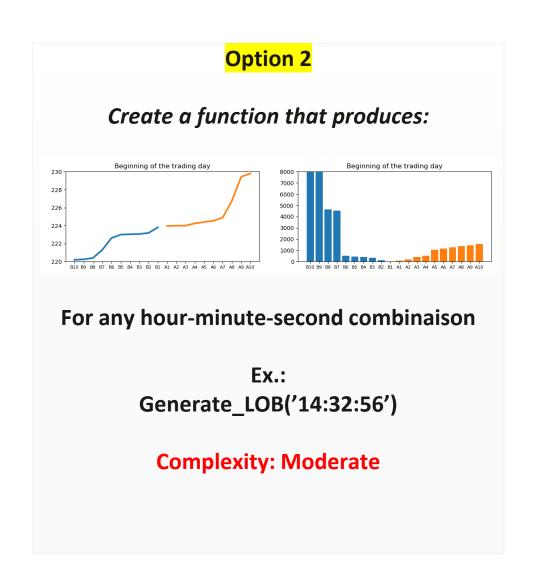
Option 1

According to you, which event type impact the most the order book?

- Submission?
- Cancellation?
- Execution?

I do not provide you a straight methodology, find our own appraoch to tackle this question!

Complexity: Advanced



PROGRAMMING EXERCISE 3 MESSAGE FILE + LOB FILE

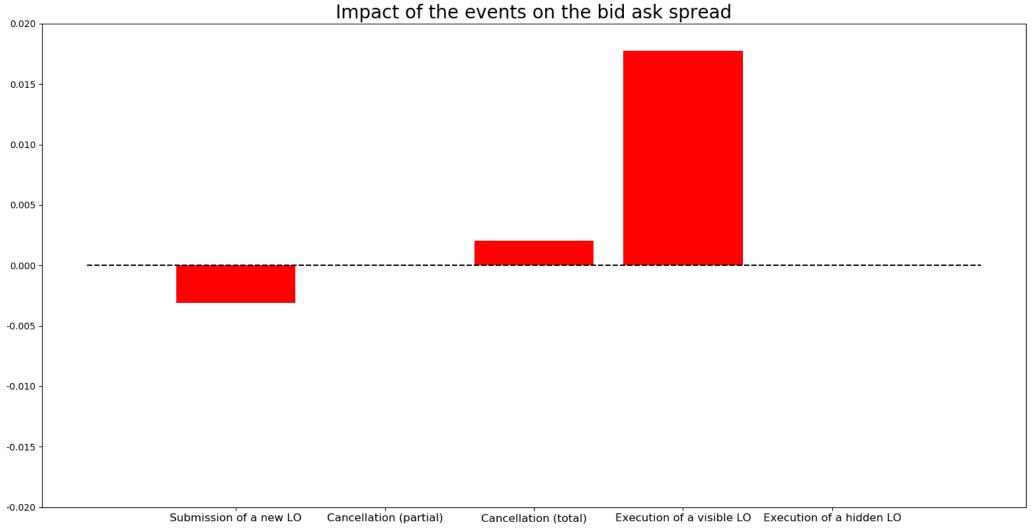


According to you, which event type impact the most the order book?

- Submission?
- Cancellation?
- Execution?

I do not provide you a straight methodology, find our own appraoch to tackle this question!





How do you explain the impact "Execution of a visible LO" on the B/A spread?

PROGRAMMING EXERCISE 4 THE TRADE FILE (1)



- Softwares
 - Suggested: Python
 - Accepted: Matlab or R....
- > TAQ data from WRDS
- File: UN_trades_full.xlsx

- 1. Date
- 2. Time
- 3. Price
- 4. Volume
- 5. Exchange Code
- Sales Condition
- 7. Correction Indicator
- 8. Sequence Number
- 9. Trade Stop Indicator
- 10. Source of Trade
- 11. MDS 127 / TRF (Trade Reporting Facility) (*)
- 12. Exclude Record

Take a look to the TAQ file pdf for column details.

PROGRAMMING EXERCISE 4 THE TRADE FILE (2)

- 1. Filtering
- 2. How much time separates two transactions?
 - i. Few seconds?
 - ii. Few minutes?
- 3. Statistics by trading hour
- 4. Statistics by stock exchange
- 5. Generate a python summary file



EXAMPLE OF PYTHON OUTPUT

Ticker: AAA

Period: XX/XX/XXXX 09:30:00 – 16:00:00

Nb. of trades: 100

Total volume: 5,222.0

First price: 46.2 Last price: 47.7

Average trade size: 50

Nb. of seconds between trades: 5

Clusters of trades: Morning/Afternoon

.....

SCREENSHOT FOR 02/05/1999



Ticker: ['UN']

Trading day: 1999-02-05 00:00:00

Time range: From 09:37:16 to

14:51:24

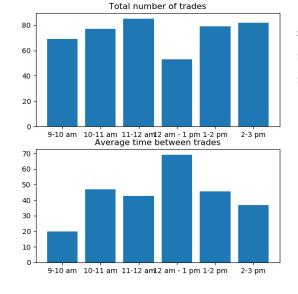
Nb. of trades: 445

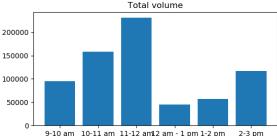
Total volume: 702600

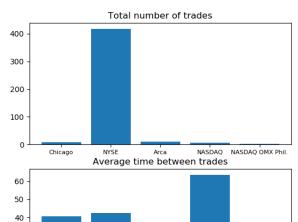
average trade size 1579

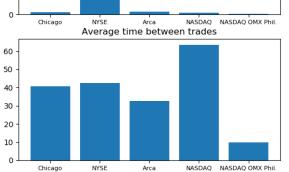
First price: 72.5 Last price: 71.75

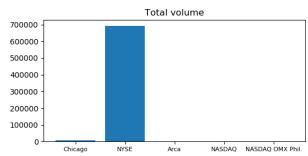
Nb. of seconds between trades: 42.0











PROGRAMMING EXERCISE 5 THE QUOTE FILE



File: UN_quotes_full.xlsx

Compute for each market-maker:

- > The best bid of the day
- > The best ask of the day
- The average bid ask spread
- More complex: average spread by hour

EXAMPLE OF PYTHON OUTPUT

Ticker: AAA

Period: XX/XX/XXXX 09:00:00 – 16:00:00

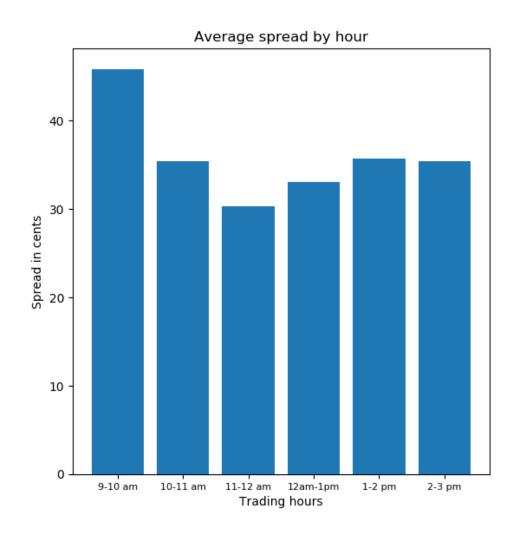
Best bid of the day: ???
Best ask of the day: ???

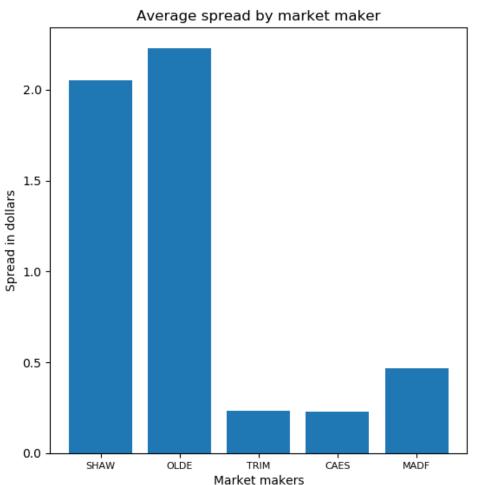
Average bid ask spread (\$): ??? Average bid-ask spread (%): ???

.

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SCREENSHOT FOR 08/03/1998 THE QUOTE FILE







Next step: Merging the trade file with the quote file.

What can we learn from that?

- ✓ Trade initiation
 - ✓ Trading cost
- ✓ Order-flow imbalance

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BEST BID AND ASK EXAMPLE 1

Market maker	Timestamp	Bid	ASk
AA	04/09/2020 10:12:03	60.3	62.3
BB	04/09/2020 10:12:04	60.5	62.5
CC	04/09/2020 10:12:05	60.0	63
Trade	04/09/2020 10:12:08	ϵ	51

✓ Best bid: 60.5

✓ Best ask: 62.3

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BEST BID AND ASK EXAMPLE 2

Makert maker	Timestamp	Bid	ASk
AA	04/09/2020 10:12:03	60.5	62.6
BB	04/09/2020 10:12:04	60.7	62.4
CC	04/09/2020 10:12:05	60.0	63
AA	04/09/2020 10:12:10	60.8	62.3
Trade	04/09/2020 10:12:10	ϵ	51

✓ Best bid: 60.7

✓ Best ask: 62.4

BEST BID AND ASK

EXAMPLE 3



Makert maker	Timestamp	Bid	ASk
AA	04/09/2020 10:12:03	60.9	62.0
BB	04/09/2020 10:12:04	60.8	62.1
CC	04/09/2020 10:12:05	60.0	63
AA	04/09/2020 10:12:10	60.7	62.1
BB	04/09/2020 10:12:11	61	61.9
CC	04/09/2020 10:12:11	61.1	61.8
Trade	04/09/2020 10:12:12		51

✓ Best bid: 60.8

✓ Best ask: 62.1

TRADE INITIATION



- I observe a transaction. Is there a way to know who has engaged the trade? Who were the most willing to trade? The buyer or the seller?
- Why is it important?
 - Computing new measures of trading costs
 - Computing order-flow imbalance
 - Quantifying informed-based trading

Some popular variables extracted from the microstructure of financial markets requires the detection of the counterparty that has engaged the trade, that is the most impatient trader.

TRADES CLASSIFICATION ALGORITHM THE TICK TEST & QUOTE TEST



Tick rule

- A trade is classified as a buy (sell) if it is executed at a price higher (lower) than that of the previous trade.
- If this is the same price: Use the closest previous price as benchmark.

Quote rule

- > Based on location of the transaction price relative to the quote midpoints.
- > If the transaction price is higher than the midpoint, a trade will be classified as a buy.
- > If a transaction price is lower than the midpoint, a trade willbe classified as a sell.
- > Trade at midpoint ==> It cannot be classified.

BUYER-INITIATED AND SELLER INITATED TRADES



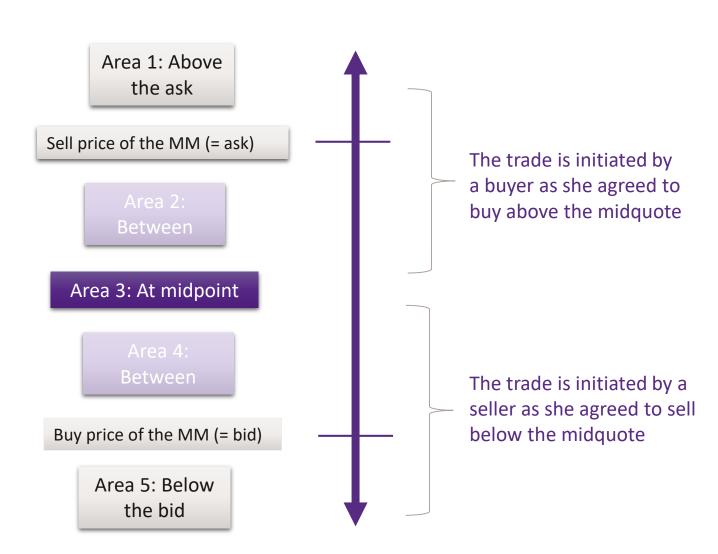
Buyer-initiated trade

THE QUOTE RULE

A trade is classified as buyerinitiated if it occurs around the ask price.

Seller-initiated trade

➤ A trade is classified as sellerinitiated if it occurs around the bid price.



MORE ADVANCED CLASSIFICATION



ALGORITHM "CLNV"

3812

B. Chakrabarty et al. | Journal of Banking & Finance 31 (2007) 3806-3821

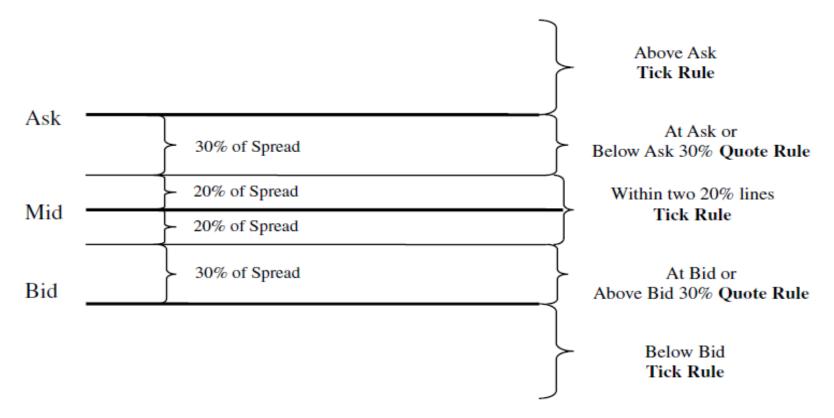


Fig. 1. Our alternative algorithm. (In order to obtain a simpler algorithm, we divided the spread into four quartiles. We use the quote rule for the two quartiles close to ask and bid and tick rule for the two quartiles close to quote midpoints. The overall accuracy rate of our alternative algorithm is 76.45% (not tabulated). One may sacrifice the improvement of accuracy of 0.07% by using this algorithm instead of the one proposed in this figure.

ORDER-FLOW IMBALANCE



Once you have classified all the trades for a trading day, you can compute the order-flow imbalance:

$$OFI = \frac{|BIT - SIT|}{\frac{BIT + SIT}{2}}$$

- BIT: daily number of buyer-initiated trades
- SIT: daily number of seller-initiated trades
- It increases around corporate events (M&A announcements, earnings announcements ...)
- It brings predictive power to explain future returns over very short horizon.

SPREADs



With the ability to classify trades, new spread can be calculated:

- 1. Effective spread
- 2. Realized spread
- 3. Total price impact
- > There is no one single formula

References:

- "Do liquidity measures measure liquidity?" by Goyenko, Holden and Trzcinka, JFE 2009.
- "Latency, Liquidity and price discovery" by Riordian and Storkenmaier, JFM 2012
- "Do prices reveal the presence of informed trading?" by Collin-Dufresne and Fos, JF 2015

QUOTED SPREAD



$$SPR_t = A_t - B_t$$

- Consider a round-trip trade: The bid ask spread is what you loose for two trades.
- The half-spread is the transaction cost per trade.
- > Empirical evidence: The wider the spread, the less liquid is the stock.

VAdvantage
Very easy to calculate

X Drawback It ignores the effect of execution inside or outside the quote

EFFECTIVE HALF SPREAD



$$EHS = 2 \times D_j \times \left[\ln(P_j) - \ln\left(MQ_j^{before}\right) \right]$$

✓ This is a measure of the execution cost actually paid by the initiator of a transaction.

Variable def	Variable definitions			
D_j	Binary variable that equals 1 for buyer-initiated orders and -1 for seller-initiated orders.			
P_j	Value-weighted average price ($P_j = \sum_i \frac{q_i}{Q} \times P_i$)			
MQ_j^{before}	A « benchmark » to select			
	 Calendar-time based benchmark ✓ Midquote point, few seconds/minutes before the transaction Transaction-time based benchmark ✓ Midquote N quotations before the transaction ✓ Alleviates differences in liquidity 			

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DECOMPOSITION OF THE EFFECTIVE HALF SPREAD

$$EHS$$

$$= 100 \times D_{j} \times \ln \left(\frac{P_{j}}{MQ_{j}^{before}} \right)$$

$$= 100 \times D_{j} \times \ln \left(\frac{P_{j}}{MQ_{j}^{before}} \times \frac{MQ_{j}^{after}}{MQ_{j}^{after}} \right)$$

$$= 100 \times D_{j} \times \ln \left(\frac{P_{j}}{MQ_{j}^{after}} \times \frac{MQ_{j}^{after}}{MQ_{j}^{before}} \right)$$

$$= 100 \times D_{j} \times \ln \left(\frac{P_{j}}{MQ_{j}^{after}} \right) + 100 \times D_{j} \times \ln \left(\frac{MQ_{j}^{after}}{MQ_{j}^{before}} \right)$$

Realized half spread

Permanent price impact

✓ Component which is subsequently reversed

✓ Component not subsequently reversed

REALIZED HALF SPREAD



$$RHS = 2 \times D_j \times \left[\ln(P_j) - \ln\left(MQ_j^{after}\right) \right]$$

- ✓ Temporary component of the effective spread
- ✓ It measures the revenue of the liquidity suppliers net of the adverse selection costs imposed by the informed on the uninformed traders.

Variable d	Variable definitions			
D_j	Binary variable that equals 1 for buyer-initiated orders and -1 for seller-initiated orders.			
P_j	Value-weighted average price ($P_j = \sum_i \frac{q_i}{Q} \times P_i$)			
MQ_j^{after}	Midquotes after the trade			
	 Calendar-time based benchmark ✓ Midquote few seconds/minutes after the transaction Transaction-time based benchmark ✓ Midquote N quotations after the transaction ✓ Alleviates differences in liquidity 			

PERMANENT PRICE IMPACT



$$PPI = 2 \times D_j \times \left[\ln \left(MQ_j^{after} \right) - \ln \left(MQ_j^{before} \right) \right]$$

- ✓ Permanent component of the effective spread.
- ✓ It measures the information content of a trade.
- ✓ New information is conveyed by the trade and it impacts the midquote permanently.
 - ✓ unlike inventory costs and other trading costs which are temporary

Variable de	Variable definitions			
D_j	Binary variable that equals 1 for buyer-initiated orders and -1 for seller-initiated orders.			
MQ_j^{before}	Midquote before the trade Few seconds before			
MQ_j^{after}	Midquote after the trade > several minutes after			



BID/ASK SPREAD WITH LIMIT-ORDER BOOK

Quoted spread	$(Ask_t - Bid_t)/(0.5 \times Mid_t)$	$Ask_t - Bid_t$
Effective spread	$D_t \times (Price_t - Mid_t) / Mid_t$	$2 \times Price_t - Mid_{t-s} $
Realized spread	$D_t \times (Price_t - Mid_{t+x}) / Mid_t$	$2 \times D_t \times (Price_t - Mid_{t+x})$
Price impact	$D_t \times (Mid_{t+x} - Mid_t) / Mid_t$	$2 \times D_t \times (Mid_{t+x} - Mid_t)$

+x: few minutes after

[&]quot;Does Algorithmic trading improve liquidity?", by Hendershoott, Jones and Mnekveld, JF 2011 "Latency, liquidity and price discovery" by Riordan and Storkenmaier, JFM 2012 "Trading your neighbor's ETFs: Competition or fragmentation" by Boehmer and Boehmer, JBF 2003

FROM TRADE LEVEL TO TIME INTERVAL



- > Spreads are computed at trade level (for each transaction).
- ➤ How do I aggregate for a given time scale?
 - Morning/afternoon/day
- Dollar-volume-weighted average.

$$\overline{SPREAD}_{i} = \sum_{k=1}^{K} \frac{q_{i,k}^{\$}}{Q_{i}^{\$}} \times SPREAD_{i,k}$$

with

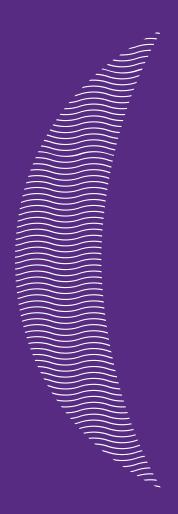
$$Q_i^{\$} = \sum_{k}^K q_{i,k}^{\$}$$

and

$$q_k^{\$} = P_k \times V_k$$

- i: interval (seconds, minutes, hours)
- k: index
- K: total number of trades in the interval i
- P: transaction price of trade k
- V: volume of the trade k in shares
- Q: volume in dollars





05

Final Project





Almost done!

Please allocate the last two hours to your final project.

Instructions



- Select one of the corporation below
- > Send an email to antoine.noel@neoma-bs.fr to book.
- First come first serve.
- > Softwares: Each final project must be done on Python (compulsory).
- Groups of 3, 4 or 5 students
- > If a one or several students in your group do not work enough, please inform the course coordinator at and a bonus/malus system will be created.

> Deadline: 06/15/2022 11.59 pm



DATASET ALLOCATION

Corporation name	Ticker	Starting date	Ending date	Group?	
Walt Disney	DIS	08/03/1998	04/30/1999		
General Electrics	GE	08/03/1998	04/30/1999		
Nike	NKE	08/03/1998	04/30/1999		
Morgan Stanley	MWD	08/03/1998	04/30/1999		
IBM	IBM	08/03/1998	04/30/1999		
Pfizer	PFE	08/03/1998	04/30/1999		
Unilever	UN	08/03/1998	04/30/1999		
Chevron	CHV	08/03/1998	04/30/1999		
ETF	QQQ	08/03/1998	04/30/1999		

(DEADLINE: 06/15/2022 11.59 pm)



Step 1

> Create your group (3-4-5 students) and select a corporation.

Step 2

- Merge the trade dataset ("Trade_File_XX.xlsx") with the quote dataset ("Quote_File_XX.xlsx") for one trading day.
 - ✓ Identify the most active MM*. Keep only its bids and asks in the quote file.
 - ✓ Choose a delay of one second to merge the trades with their prevailing bid ask pair.
 - ✓ Python function for merging: pandas.merge_asof
 - ✓ https://pandas.pydata.org/pandas-
 docs/version/0.25.0/reference/api/pandas.merge
 association
 docs/version/0.25.0/reference/api/pandas.merge
 association
 https://pandas.merge
 association
 docs/version/0.25.0/reference/api/pandas.merge
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^{*:} largest amount of bid and ask updates.

(DEADLINE: 06/15/2022 11.59 pm)



Step 3

Now, for each transaction, you have now a prevailing bid and a prevailing ask. You can now classify trades. To do so, please implement the CLNV algorithm (slide 120).

Step 4

> Compute the proportion of trades inside quotes, outside quotes and at quotes.

Step 5

Compute the daily <u>weighted</u> average effective spread.

(DEADLINE: 06/15/2022 11.59 pm)



Step 6

> Repeat step 2 to step 5 for each trading day of your dataset.

Step 7

> Generate the effective spread and order-flow imbalance time series. Show these two time series on a single plot.

Step 8 (BONUS)

- For each transaction, all prevailing bids and asks are not the ones of the most active MM.
- Rather, keep all market makers and please select the highest bid and the lowest ask among the most contemporaneous bid / ask pair of each market maker (slide 116).