

WRDS E-Learning Session Overview of TAQ on WRDS

Insights on Efficient Programming in Intraday Research

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E-Learning Session Guidelines

Welcome to WRDS E-Learning. Please use the following guidelines:

- Ask Questions via the "Chat" Window.
- Audio problems? Disconnect and call in using one of the phone numbers listed in the reminder email.

Agenda

Introduction to TAQ and its research applications

Efficient Programming with TAQ datasets

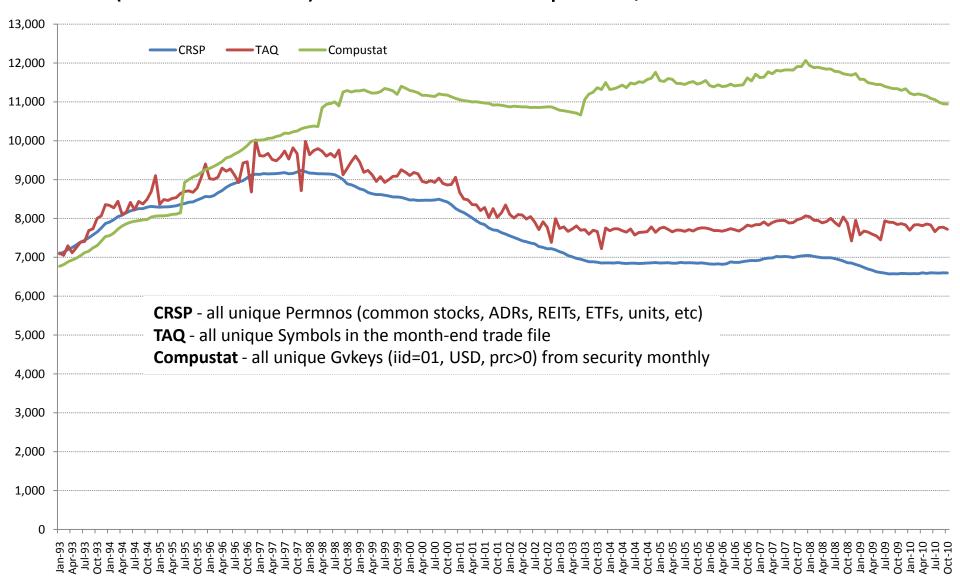
The Lee-Ready Algorithm

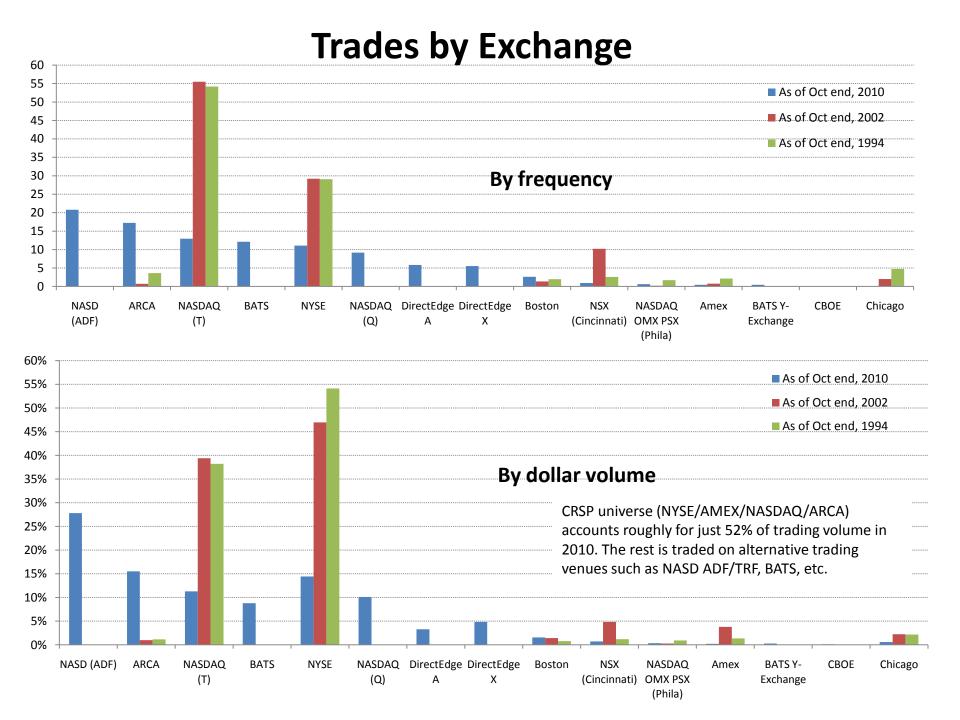
Generating NBBO with TAQ Quotes

What is TAQ?

quo	des and Quotes Database contains intraday tick-by-tick transactions data (trades and tes) for all securities (not aggregate indexes) listed on the NYSE and AMEX as well as daq National Market System (NMS) and SmallCap issues back to 1993
	Excludes transactions reported outside of the Consolidated Tape hours of operation (8.00 am-6.30 pm) Excludes trading in NYSE-listed securities between 8-9.30 am by other markets
on 2	of the latest available vintage (Oct, 2010) TAQ covers more than 7,700 securities traded 15 different trading venues. The package includes Consolidated Trades& Quotes (timenped up to a whole second), Master and Dividend datasets
	Common stock -81%, Derivatively priced (e.g., ETFs) – 15.3%, Preferred – 3.3%, Warrant-0.5%, Right-0.03%, other- 0.02% (oct 2010).
	TAQ does NOT carry intraday data for either derivatives or OTC securities. OPRA database provides intraday last sale information and options quotations from a committee of Participant Exchanges, whereas Compustat Security file contains daily data for OTC stocks, among others.
	of trades, and especially quotes, datasets expanded exponentially during the last few rs, to a large extent due to the proliferation of algorithmic trading
	The use of algorithmic trading has grown considerably and, as of 2007, was estimated to account for one-third of equity trading volume in the United States (Hendershott, Jones, and Menkveld, 2010)
	A handful of HFT firms (2 percent of the 20,000 trading firms in the US) accounted for an estimated 70% of overall trading volume on US equities markets in 2009 (WSJ, Jul 9, 2009)

Number of distinct traded securities: TAQ (end of month) vs. CRSP vs. Compustat, Jan 1993-Oct 2010





Trading Venues for US equities, Jan-Jun 2008

ECN- Electronic Communication Network ATS-Alternative Trading System

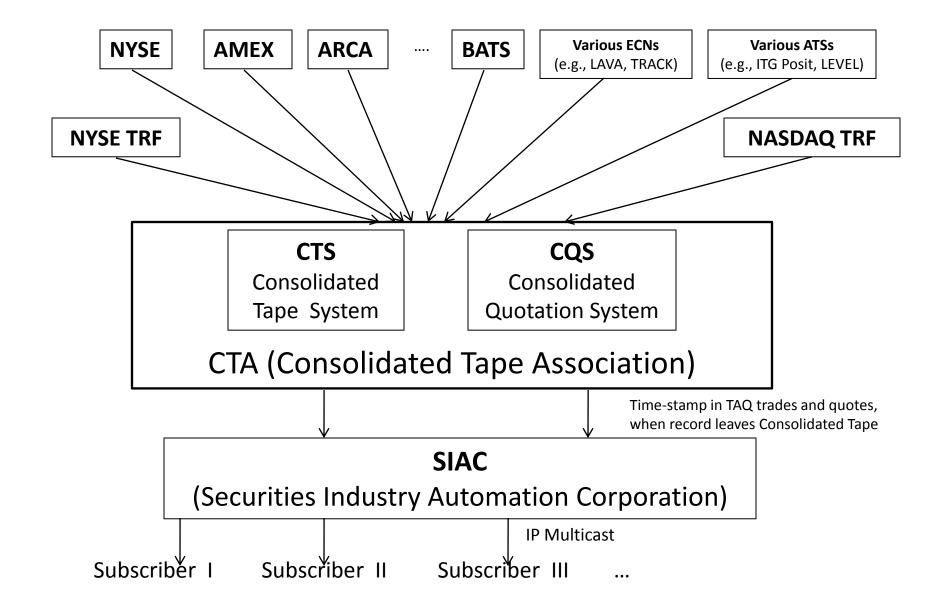
EXCHANGES	ECNS	A	TS
NASDAQ	BATS	ITG POSIT	CITIMATCH
NEW YORK STOCK EXCHANGE	DIRECTEDGE	BIDS	CS CROSSFINDER
ARCA PELAGO	TRADEBOOK	LEVEL	LX
NATIONAL STOCK EXCHANGE	LAVA	LIQUIDNET	MLXN
AMERICAN STOCK EXCHANGE	TRACK	MATCHPOINT	SIGMA X
CHICAGO STOCK EXCHANGE		INSTINCT	MORGAN STANLEY POOL
PHILADELPHIA STOCK EXCHANGE		MILLENNIUM	UBS PIN
BOSTON STOCK EXCHANGE		PIPELINE	BNY CONVERGEX
INTERNATIONAL STOCK EXCHANGE		PULSE	FIDELITY CROSS STREAM
CHICAGO BOARD OPTIONS EXCHANGE		ESPEED AQUA	LAVA ATS

Source: O'Hara and Ye (2010), period Jan-Jun 2008

Regulation (Reg) NMS and Trade Reporting Facilities (TRFs)

- Reg NMS (2005, http://www.sec.gov/rules/final/34-51808.pdf) required traders to transact on a trading venue at the lowest price rather than on a venue offering the quickest execution or the most reliability. Imposed caps on access charges that exchanges and other venues could levy > explosion of trading venues (>40 platforms in 2008)
- Before the SEC mandate off-exchange volume was aggregated with exchangeexecuted volume for reporting purposes. After March 5, 2007 exchanges must report only their on-exchange volumes, with off-exchange volumes handled by TRFs that report directly to the consolidated tape
 - TRF data in TAQ does not disaggregate trades into specific execution venues. So one can only
 determine the aggregate off-exchange volume per security, rather than specific volume of trading in
 each of the many non-exchange venues.
- Currently, the NYSE TRF, the Nasdaq TRF, and the National Stock Exchange (NSX)
 TRF are active. In addition, the Alternative Trade Facility (ADF) also operated as a
 TRF.
 - The ADF was originally created by the NASD in response to the Nasdaq market's conversion to forprofit status. The ADF includes both a reporting and display facility, allowing trading platforms who do not wish to post quotes on the Nasdaq an alternative venue in which to display quote and trade information (O'Hara and Ye, 2010)

Overview of Trade and Quote Reporting Procedures



Research applications using intraday data

- Signing the order flow to identify buyer- and seller-initiated trades
 - 5-second rule (Lee and Ready, 1991)
 - Revised algorithms for matching trades and quotes (Vergote, 2005; Rosenthal, 2008)
 - Issues with Lee and Ready when identifying short sales (Asquith, Oman and Safaya, 2010; Chakrabarty et al., 2010)
- Estimation of Transaction Costs, Liquidity and Adverse Selection
 - Validation of Bayesian Gibbs estimates of effective cost using TAQ (Hasbrouck, 2010)
 - Various versions of Bid-Ask spread: quoted, effective and realized spreads and price impact (Goeynko, Holden and Trzcinka, 2009; Brennan et al., 2009), high-low spread estimator (Corwin and Schultz, 2010), VWAP (Hu, 2009)

Investor Behavior

- Institutional trading (e.g., Campbell, Ramodarai and Schwartz (JFE, 2009) infer daily institutional trading behavior from TAQ using a sophisticated method that best predicts 13-F data from trades of different size)
- Retail trading (e.g., Barber, Odean and Zhu, 2006; small trades of Hvidkjaer, 2008)

Research applications using intraday data (cont'd)

- Measures of uncertainty and information risk
 - PIN (Easley, Hvidkjaer, O'Hara, 2002)
 - Flow Toxicity/VPIN (Easley, Lopez de Prado, O'Hara, 2010, 2011)
- Impact of decimalization (before and after)
 - Cost of trading and market quality (Bessembinder, 2003)
 - Stock return implications (Graham, Michaely and Roberts, 2003)
- Tests of market efficiency
 - Speed of price discovery, reaction to order flows and imbalances (Chordia, Roll and Subrahmanyam, 2005&2008)
 - Predictable periodicity in intraday returns (Heston, Korajczyk, Sadka, 2010)
- High-Frequency and Algorithmic Trading
 - Impact of HFT on market quality (Brogaard, 2010), Low-latency (Hasbrouck et al., 2011)
 - Floor trader vs automation (Venkataraman, 2001; Cardella, Hao, Kalcheva, 2010)
 - Transparency of the trading process (Bessembinder, Panayides and Venkataraman, 2009)

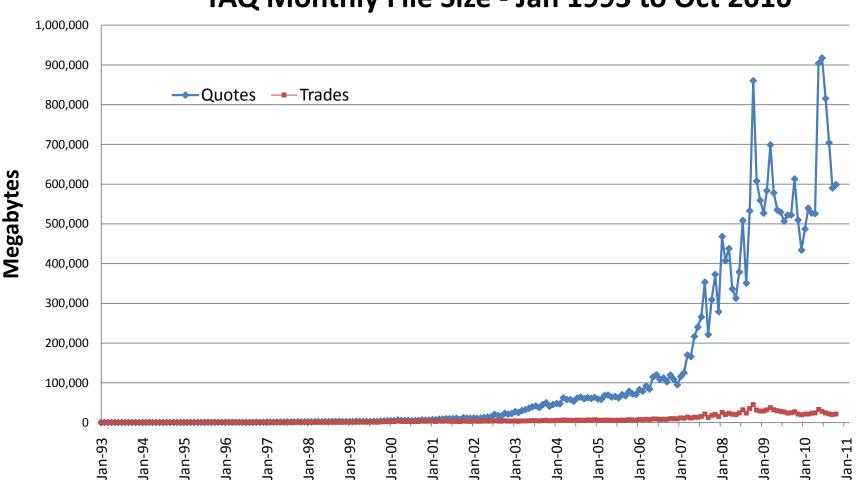
Efficient Programming with the TAQ Database

TAQ Data Organization at WRDS

- Stored as SAS datasets
- Located in /wrds/taq/sasdata
- Covers all trades & quotes since Jan 1993
- Occupies about 32 Terabytes
- Typical recent month adds 850 gigabytes

Major Feature of TAQ Data: Size

TAQ Monthly File Size - Jan 1993 to Oct 2010



Four Types of TAQ Data Files

- CQ: Consolidated Quotes
 - One record per quote (468 billion for Jan 1993-Oct 2010)
- CT: Consolidated Trades
 - One record per trade (39 billion for Jan 1993-Oct 2010)
- MAST: Master
 - Tracks the CUSIP and Exchanges for each SYMBOL
- DIV: Dividends
 - Provides Dividends by Date and SYMBOL

CT and CQ datasets "Consolidated" Trades and Quotes

- "Consolidated"

 Refers to centralization of record processing

 NOT to aggregation of trades or quotes.
- One dataset per trading date.
- Named as CX_yyyymmdd:
 CT_19930104 covers trades on Jan 4, 1993
 CQ 20101001 covers quotes on Oct 1, 2010

CT and CQ datasets Large Size and Tied TIME Values

- Each CQ/CT dataset indexed by SYMBOL
 - Support faster filtering by SYMBOL
- Each dataset sorted by SYMBOL TIME
 - Supports "BY processing" (see WRDS SAS e-learning)
- TIME is in whole seconds, yielding many ties:
 - CT files: 5% (1993) to 88% (2010)
 - CQ files: 25% (1993) to 98% (2010)
- Records with tied TIME from data supplier are not reordered in WRDS datasets.

CT Dataset Contents

One Record per Trade

Sample Contents from CT 20100101

0bs	SYMBOL	DATE	TIME	PRICE	SIZE	G127	CORR	COND	EX
10001	^	00101001	10.07.00	00 57	100	0	0	0	NI.
10001	Α	20101001	10:27:38	33.57	100	0	0	a	N
10002	Α	20101001	10:27:38	33.57	100	0	0	a	Р
22000001	SS0	20101001	10:10:58	39.75	100	0	0	F	В
22000002	SS0	20101001	10:10:59	39.76	1255	0	0	@	D

CQ Dataset Contents

One Record per Quote

Sample Contents from CQ 20100101

OBS	SYMBOL	DATE	TIME	BID	OFR	BIDSIZ	OFRSIZ	MODE	EX	MMID
	_				_					
100001	Α	20101001	9:39:40	0.00	0	0	0	12	С	
100002	Α	20101001	9:39:40	33.88	0	6	0	12	C	
52000001	US0	20101001	10:46:45	35.21	35.22	4	107	12	Т	
520000002	US0	20101001	10:46:45	35.21	35.22	12	68	12	Z	

Common TAQ Programming Issues

- Large TAQ datasets engender excessive resource use
 - Disk I/O (input/output)
 - Disk storage space for intermediate datasets
 - LONG TIME TO COMPLETION

- The Primary Pitfalls of User Programs are:
 - EXCESSIVE USE OF PROC SORT
 - Inefficient extraction of subsets
 - Generating unneeded intermediate datasets
 - Using PROC SQL to sort TAQ data

Listing Multiple Datasets With One Name Prefix

The SET statement accepts name prefix.
 (a partial dataset name truncated by a ":")

 Read all October 2010 trade datasets with: set taq.ct_201010:;

Other examples (don't try these):

```
set taq.cq_:; ** All TAQ quotes **; set taq.ct_2005:; ** All 2005 trades **;
```

Sample Data Extraction

```
libname taq '/wrds/taq/sasdata';
data mydata;
   set taq.cq 201010: ;
   where symbol in ("IBM", "DELL", "MSFT", "CSCO");
   if bid>0 and ofr>bid and (ofr-bid)/(ofr+bid) < .1;
   spread=ofr-bid;
run;
proc means data=mydata n min max mean;
 class symbol;
 var spread;
run;
```

This program took 2 Minutes, 16 Seconds

Replace intermediate datasets with Views

Making a data VIEW does not write to disk. It "pipes" the data directly to the PROC MEANS

```
data v_mydata / view=v_mydata;
    set taq.cq_201010: ;
    where symbol in ("IBM","DELL","MSFT","CSCO");
    if bid>0 and ofr>bid and (ofr-bid)/(ofr+bid) < .1;
    spread=ofr-bid;
run;
proc means data=v_mydata n min max mean;
    class symbol;
    var spread;
run;</pre>
```

This program took 1 Minute, 12 Seconds

Time Filters with TAQ Data

You can use "Time Literals" to filter the TIME variable

```
data vmydata / view=vmydata;
  set taq.cq_20101001;
  where symbol in ('IBM','DELL')
     and time between '09:30:00't and '16:00:00't;
     ... other statements here ...;
run;
proc means data=vmydata;
    ...
run;
```

Sorting by DATE/SYMBOL/TIME IS SUPERFLUOUS

This SET statement reads all datasets with prefix CQ_201010 in alphabetic order (same as date order)

```
data vtemp / view=vtemp;
  set taq.cq_201010: ;
  where symbol in ('IBM','DELL');
  run;
```

SET reads all the cq_201010xx datasets consecutively, so VTEMP is already sorted by DATE SYMBOL TIME. As a result ...

THIS SORT IS SUPERFLUOUS

```
proc sort data=vtemp out=temp; ** NOT NEEDED **;
  by date symbol time;
run;
```

"OPEN=DEFER" Saves Memory

Accessing multiple daily datasets (instead of one monthly dataset) uses more memory. SAS assigns a memory buffer for each dataset read by the SET statement.

```
data mydata;
  set taq.cq_201010: ;
  where symbol in ('IBM','DELL');
run;

Memory Use: 573K
```

Using "open=defer" tells SAS to "reuse" the same memory buffer for each incoming data set

```
data mydata;
  set taq.cq_201010: open=defer;
  where symbol in ('IBM','DELL');
run;
```

Memory Use: 339K

Ordering by SYMBOL/DATE/TIME The Naïve Sort

This SET reads datasets in order, yielding DATE/SYMBOL/TIME order.

data vtemp /view=vtemp;
 set taq.cq_201010: ;
 where symbol in ('IBM','DELL');
 run;

To get SYMBOL/DATE/TIME order, a PROC SORT is needed

```
proc sort data=vtemp out=mydata;
  by symbol date time;
run;
```

Note: In PROC SORT, records with tied TIME values are NOT reordered

Ordering by SYMBOL/DATE/TIME Avoiding the Naïve Sort

Using a BY statement with the SET statement will allow the incoming data to be "interleaved". I.e. all the "A" SYMBOLs are processed first, then the "B" SYMBOLs, etc.

```
data mydata;
  set taq.cq_201010: ;
  by symbol date time;
  where symbol in ('IBM','DELL');
run;
```

Because the incoming datasets have a constant DATE value, and are sorted by SYMBOL/TIME, the datasets can be read "in parallel" to produce the desired order, without using PROC SORT.

PROC SORT vs. "order by" (PROC SQL)

 PROC SORT guarantees order preservation for tied records.

- The PROC SQL "order by" clause does not guarantee order preservation for tied records (e.g. "order by symbol, date, time")
- WRDS Strongly Recommends using PROC SORT for ordering TAQ data rather than PROC SQL

SAS and TAQ Links at WRDS

SAS E-learning links on:

http://wrds-web.wharton.upenn.edu/wrds/E-Learning/

"SAS at WRDS"

"Intermediate SAS at WRDS"

Other comments on Using TAQ Efficiently:

http://wrds-

web.wharton.upenn.edu/wrds/support/Data/ 005Using%20TAQ%20Data %20Efficiently/index.cfm

PROC SQL "ORDER BY" with tied records:

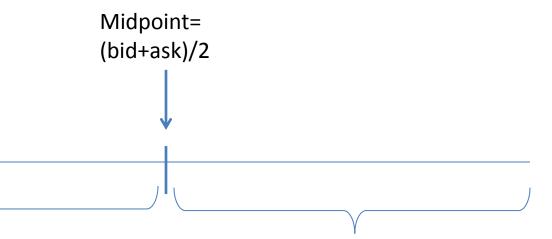
http://wrds-

web.wharton.upenn.edu/wrds/support/Data/ 005Using%20TAQ%2 0Data%20Efficiently/ 017Note on Usage of SQL with TAQ data. cfm

Lee and Ready (1991) Algorithm

Who initiated a trade?





Seller-initiated if Price < Midpoint

Buyer-initiated if Price > Midpoint

2. Tick test if price=midpoint

-**Seller**-initiated: P(t) < P(t-1)

-Buyer-initiated : P(t)>P(t-1)

(in case of equality, the procedure can go back several lags).

Are 'Trades' and 'Quotes' aligned?

Delay of the reported trade time wrt a quote time

- There are several assumptions about misalignment between trade time and reported quote time
- Lee-Ready (1991), Peterson and Sirri (2003),
 Henker and Wang (2005)
- Lee-Ready for example use a "5-second" difference between trade time and quote time.

WRDS-sample program

Main Characteristics and Efficiency Issues

Lee and Ready Sample SAS program:

http://wrdsweb.wharton.upenn.edu/wrds/support/Data/ 004Research%20Applications/ 010Market%20Microstructure%20S tudy/lee_ready.cfm

- Use SAS Views.
- Use 5-second rule (can be easily modified).
- Multiple entries by second: assumptions needed. Currently, no millisecond time variable=> how to merge quotes and trades?
- Take advantage that TAQ files are sorted by SYMBOL and TIME.

```
data lee_ready;
set quote_file trade_file;
by date symbol time;
...
```

Output-sample

Stock Symbol	Quote date	Quote time	Price	mid_point_B id_Ask	lagprice	lag2price	TICK Test	buyer_init	
DELL	20090120	9:45:01	10.177	10.185	10.18	10.19	-1	1	seller
DELL	20090120	9:45:29	10.164	10.160	10.16	10.1607	1	. 1	buyer
DELL	20090120	9:46:47	10.125	10.125	10.13	10.1254	-1	-1	seller
DELL	20090120	9:48:23	10.085	10.085	10.08	10.0804	1	. 1	buyer
DELL	20090120	9:49:24	10.08	10.08	10.08	10.0888	-1	-1	seller
DELL	20090120	9:54:39	10.02	10.02	10.02	10.02	0	0	not defined

Generating NBBO From TAQ Quotes

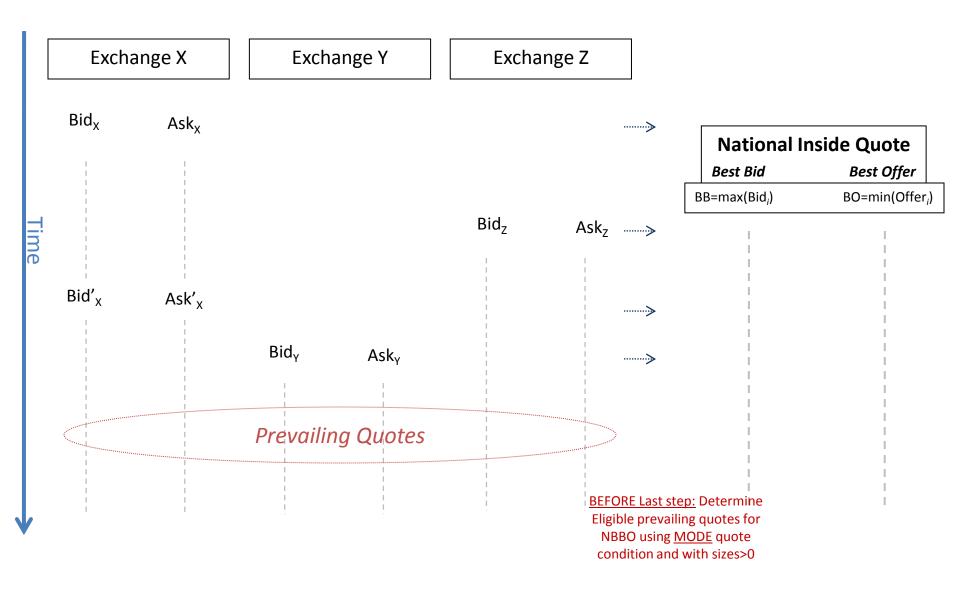
NBBO Research Application

- Regulation NMS National Market System: calculate National Inside Quote from valid quotes across various exchanges: Best Bid (BB), Best Offer (BO), Size at BB, Size at BO
- Quote Rule: http://www.sec.gov/answers/trdexbd.htm
 - best price at which a market maker is willing to trade
 - a prevailing quote of an exchange is the most recent valid quote from that exchange
 - every new quote by an exchange supersedes its previous prevailing quote
- NBBO calculation every second because of TAQ Monthly feed time stamp granularity
- Evaluation of accuracy of WRDS NBBO algorithm:
 http://wrds-web.wharton.upenn.edu/wrds/support/Data/ 005Using%20TAQ%20Data%20Efficiently/ 028NBBO note.cfm
 - Minimize the percentage of locks and crosses within an NBBO
 - Maximize the percentage of Inside Trades
- SAS Code: http://wrds-web.wharton.upenn.edu/wrds/support/Data/ 005Using%20TAQ%20Data%20Efficiently/ 028NBBO.cfm
- Invocation: %nbbo (yyyymmdd = 20080606 , outset = myfile);
- Evaluation of execution speed of WRDS NBBO algorithm SAS Macro on WRDS server
 - To create a 1998 NBBO daily dataset for all Stocks: 26 seconds
 - To create a 2006 NBBO daily dataset for all Stocks: 9 minutes 37 seconds

NBBO Algorithm: TAQ Quote Tape

	Stock Symbol	Time	Bid	Ask	Exchange	Mode (quote condition)
	ABC	11:30:02	Bid _X (1)	Ask _x (1)	x	12
+	ABC	11:30:05	Bid _z (2)	Ask _z (2)	Z	12
T: 33	ABC	11:30:06	Bid _y (3)	Ask _y (3)	Y	12
	ABC	11:30:08	Bid _x (4)	Ask _x (4)	X	
	ABC	11:30:40	Bid _y (5)	Ask _y (5)	Y	10
	ABC	11:31:22	Bid _Y (6)	Ask _y (6)	Y	12
	ABC	11:32:50	Bid _z (7)	Ask _z (7)	Z	12

NBBO Algorithm: SAS Code Logic



NBBO Output Example: IBM

Stock Symbol	Quote date	Quote time	Best Bid	Best Offer	Best Bid Size	Best Offer Size	Total Number of New Quotes	# of Exchanges with Valid Prevailing Quotes
IBM	19970303	9:47:13	144.625	144.875	36	513	1	7
IBM	19970303	9:47:16	144.625	144.875	36	508	1	7
IBM	19970303	9:47:34	144.75	144.875	50	508	1	7
IBM	19970303	9:47:35	144.75	144.875	50	508	1	7
IBM	19970303	9:47:36	144.75	144.875	51	508	1	7
IBM	19970303	9:47:37	144.75	144.875	51	508	2	7
IBM	19970303	9:47:38	144.75	144.875	51	508	2	7
IBM	19970303	9:47:39	144.75	144.875	101	513	3	7
IBM	19970303	9:47:40	144.75	144.875	101	513	3	7
IBM	19970303	9:47:45	144.75	144.875	101	513	1	7
IBM	19970303	9:48:07	144.75	144.875	151	513	1	7
IBM	19970303	9:48:20	144.75	144.875	151	613	1	7
IBM	19970303	9:48:27	144.75	144.875	166	608	1	7
IBM	19970303	9:48:56	144.75	144.875	16	608	1	7
IBM	19970303	9:48:58	144.75	144.875	15	608	1	7
IBM	19970303	9:49:00	144.75	144.875	15	608	4	7
IBM	19970303	9:49:01	144.75	144.875	10	608	1	7
IBM	19970303	9:49:05	144.625	144.875	60	613	2	7
IBM	19970303	9:49:06	144.75	144.875	50	613	1	7

Future of Microstructure Research on WRDS

- WRDS High-Performance TAQ Platform (WRDS HPP)
 - The new WRDS HPP is optimized for TAQ access while the current WRDS server is a general-use server
 - Users on WRDS HPP will be able to process ALL TAQ data in days when it could take weeks otherwise
 - The new WRDS HPP contains TAQ millisecond data, as well as TAQ monthly feed data, and will include TAQ-CRSP linking tables, and NBBO datasets in SAS format
 - WRDS High-Performance TAQ Platform is available at an additional cost. Please contact WRDS-Support@wharton.upenn.edu for more information
- Research Applications Work-in-Progress
 - TAQ-CRSP Linking Macro: TCLINK.SAS that maps PERMNO to TAQ Symbol
 - Buy-Sell Imbalance Research Application
 - Intraday Return and Intraday Event Study Research Application

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