

# Algorithmic Trading Strategies<sup>1</sup>

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<sup>1</sup>Reference: Avellaneda (2011), Maglaras (2015)

# Algo Trading Systems: Typically Decomposed into 3 Steps

- **Trade scheduling (macro-trader):** splits parent order into  $\sim 5$  min slices (**Lecture 2**)
  - Relevant time-scale: minutes-hours
  - Schedule follows user selected strategy (VWAP, POV, IS, ... )
  - Reflects urgency, alpha, risk/return tradeoff
  - Schedule updated during execution to reflect price/liquidity/...
- **Optimal execution of a slice (micro-trader):** further divides slice into child orders (**Lecture 3**)
  - Relevant time-scale: secondsminutes
  - Strategy optimizes pricing and placing of orders in the LOB
  - Execution adjusts to speed of LOB dynamics, price momentum, ...
- **Order routing:** decides where to send each child order (**Lecture 4**)
  - Relevant time-scale:  $\sim 1 - 50$  ms
  - Optimizes fee/rebate tradeoff, liquidity/price, latency, etc

# Algo Trading Strategies: VWAP

- **VWAP (Volume-Weighted Average Price):** Trades according to forecasted volume profile to achieve (or beat) the market VWAP.
  - <sup>2</sup> If an asset during some time interval has  $N$  trades with price  $p_k$  and volume  $v_k$ , its VWAP is

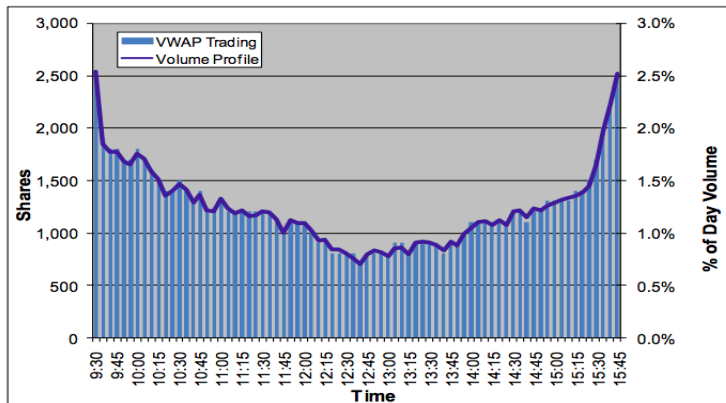
$$VWAP = \sum_{k=1}^N v_k p_k / \sum_{k=1}^N v_k$$

- Passive strategy.
  - Subject to significant market risk (why?).
- Algorithm:
  - Estimate the average volume traded in every 5 minute interval using historical data.
  - In each time-interval, execute an amount proportional to the normative volume for that interval.

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<sup>2</sup>Yang (n.d.)

# Volume-Weighted Average Price



- Properties:

- The algorithm always concludes (trade sizes are known in advance).
- Volume function is estimated using historical data. This may not correspond exactly to *ex-post* VWAP.

# Algo Trading Strategies: TWAP

- **TWAP (Time-Weighted Average Price):** Trades uniformly over time to achieve (or beat) TWAP benchmark.
  - Passive strategy
  - Also subject to market risk
- <sup>3</sup> Such a simple protocol has a risk of exposure of the traders intentions to other market participants:
  - *For example, some scalpers may realize that a large order is being traded and start trading the same instrument in expectation that the large trading volume will inevitably move the price.*
- To prevent information leak, TWAP schedule may be randomized in terms of size and submission time of child orders.
  - *For example, if the trading interval is four hours, 25% of the trading volume must be executed each hour, and the child order size may be adjusted deterministically for each hour.*

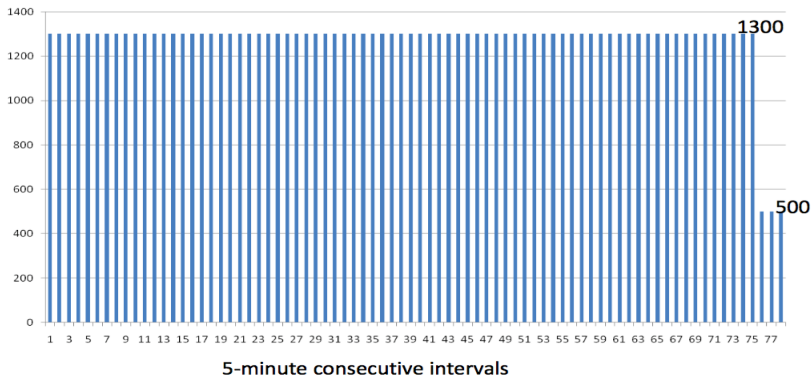
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<sup>3</sup>Yang (n.d.)

# Time-Weighted Average Price

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Figure 1: Example: 100,000 shares TWAP/all day



- Equal amount of shares in each period of time.
- Not very popular in practice.

# VWAP vs. TWAP

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- During a slow trading day, the TWAP may be very similar to the VWAP, even to the penny at times. However, in a volatile session, or when volume is higher than usual, the two indicators may diverge (how and why?).

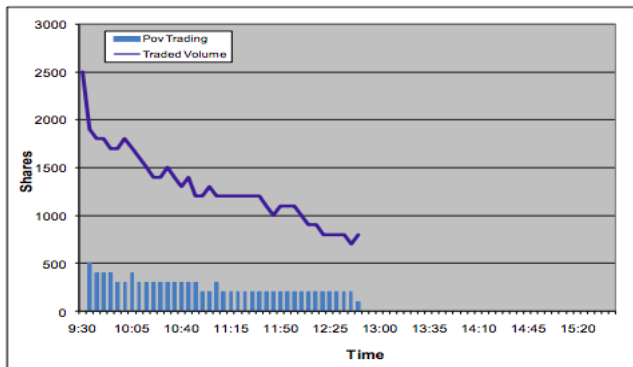
<sup>5</sup>Yang (n.d.)

- **POV (Percent of Volume):** Submit child orders with sizes equal to a certain percentage of the *total trading volume*.
  - Execute while tracking the realized volume profile at a constant target participation rate, e.g., buy IBM at 15% participation rate.
  - Controls behavior during volume spikes to avoid excessive cost.
  - Popular in practice with  $\sim 5\% - 30\%$  participation rates.
  - Participation rate is highly related to transaction cost (why?).



# Price of Volume

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- The POV (Percentage of Volume) algorithm addresses the problem of VWAP by using the actual total traded volume of the day as benchmark.

<sup>6</sup>Avellaneda (2011)

- **IS (Implementation Shortfall):** Schedules trade so as to optimally tradeoff *expected shortfall* (cost) against *execution risk*.
  - Objective functions is usually a weighted linear sum of execution cost and execution risk.
  - Execution speed adapts with respect to changes in market conditions.
  - Popular, especially with portfolios with intricate cost/risk tradeoff.

# Implementation Shortfall

- (Almgren and Chriss, 2000) **Expected Fall**

- Stock price is subject to price impact.

$$\tilde{S}_t = S_t - \eta \frac{dQ(t)}{dt}.$$

- *Expected execution cost* is the volume-weighted execution price along the time:

$$E = -\mathbb{E}\left[\int_{t=0}^T \tilde{S}_t \frac{dQ(t)}{dt} dt\right].$$

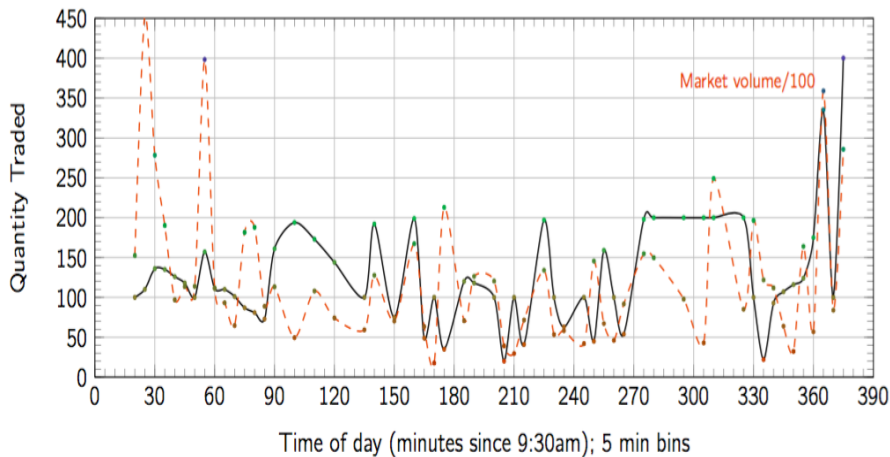
- *Execution risk* is the variance of execution position:

$$V = \text{Var}\left[\int_{t=0}^T Q(t) dS_t\right]$$

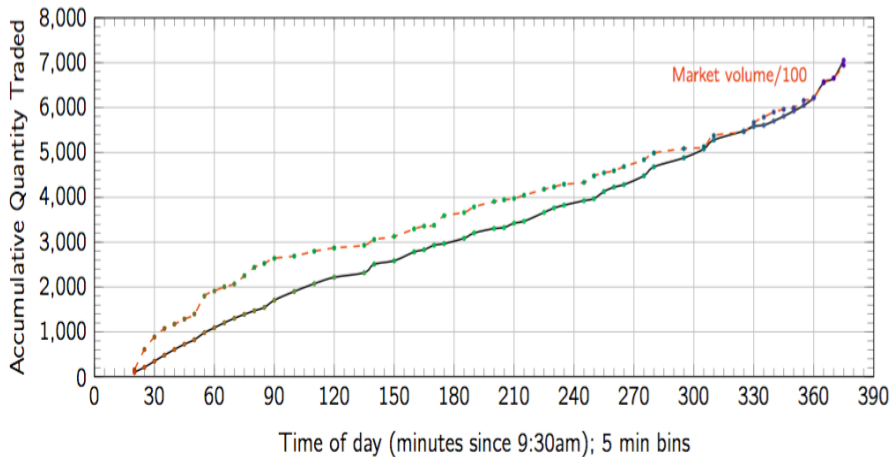
- The optimal position is the one that minimizes the sum of *execution cost* and *execution risk*:

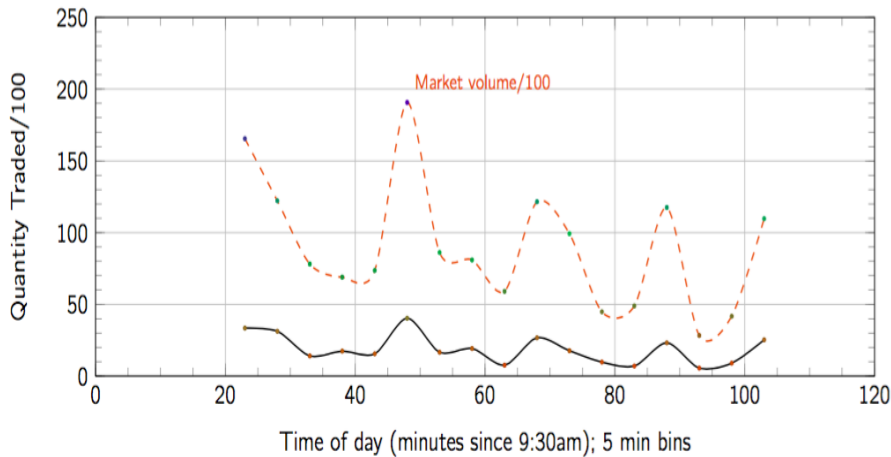
$$\min_Q \{E + \lambda V\}$$

# VWAP, XLY, 07/22/2013 ( .15%ADV )

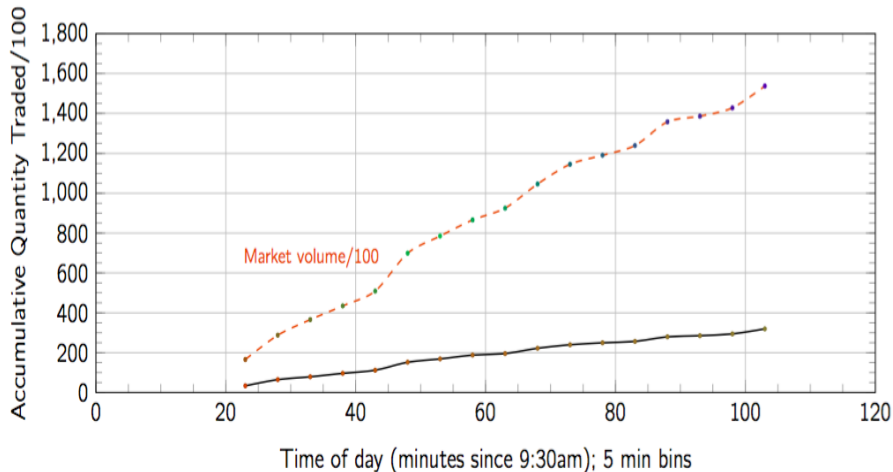


# VWAP, XLY, 07/22/2013 (cumulative quantity)

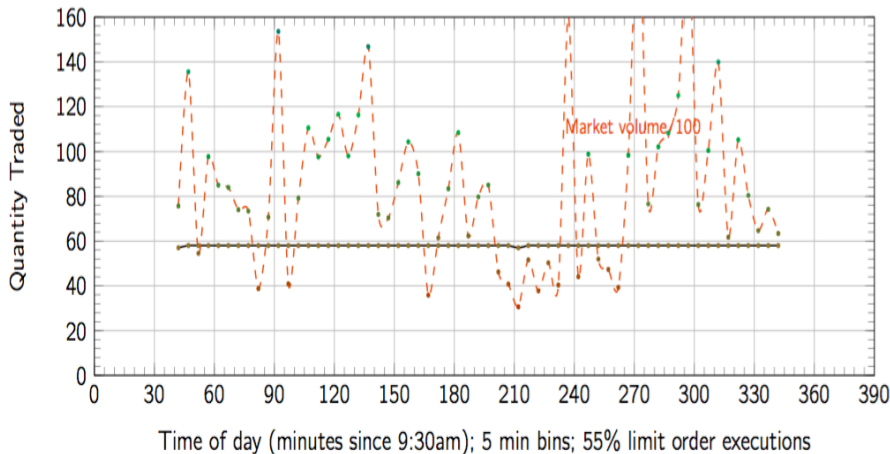




# POV, ACT, 07/08/2013 (cumulative quantity)

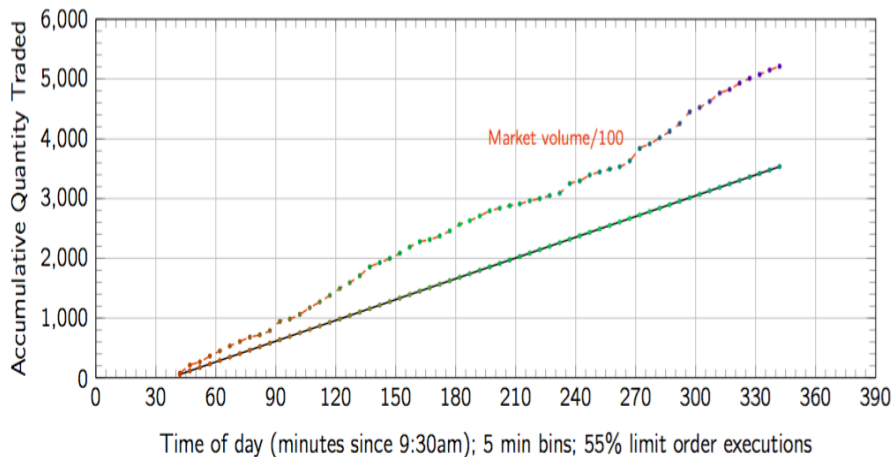


# Schematic of execution profiles: TWAP, XLY, 07/02/2013





# TWAP, XLY, 07/02/2013 (cumulative quantity)



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