## Optimal estimation of intraday 'true value' using the Glosten-Milgrom model Isaac Drachman

We wish to statistically model the 'true value' of a traded asset over a time period solely by utilizing its order flow. The primary object of study is therefore  $\mu_t$ , the true value after the t-th trade. We say that  $\mu_t$  follows a distribution  $F_t$ . We let our best estimate be  $\hat{\mu}_t = \mathbb{E}F_t$ .

We consider the data  $(m_0, X)$ , where  $m_0$  is the initial midpoint price and X is a sequence of trades  $x_1, x_2, ..., x_T$ . The t-th trade  $x_t$  is the pair  $(d_t, p_t)$ , where  $d_t = +1$  or -1 for a market buy or a market sell, respectively, and  $p_t$  is price at which the trade occurred. We specify the first prior distribution  $F_0$  as Normal with mean  $m_0$  and std. deviation  $\sigma$ , which will be a parameter of our model. Our first best estimate is  $\hat{\mu}_0 = m_0$ .

From the Glosten-Milgrom microstructure model, we say that all traders are either perfectly informed (who know  $\mu_t$ ) or uninformed (noisy trading) with the following specifications (where  $b_t$ ,  $a_t$  are bid/ask after trade t)

$$P(\text{trade made by informed trader}) = \alpha$$

$$P(\mu_t > a_t \mid \text{informed bought}) = 1$$

$$P(\mu_t < b_t \mid \text{informed sold}) = 1$$

$$P(\text{uninformed buys when trades}) = \eta$$

After witnessing trade  $x_t$  we update  $F_t$  to a posterior distribution  $F_{t+1}$  using this<sup>[1]</sup>. We explicitly modify the probability distribution function for a trade with  $d_t = +1$  (i.e. market buy)

$$f_{t+1}(p) = f_t(p) \times \begin{cases} (1-\alpha)\eta + \alpha & p > a_t \\ 1-\alpha - (1-\alpha)\eta & p \le a_t \end{cases}$$

After witnessing a trade with  $d_t = -1$  (i.e. market sell)

$$f_{t+1}(p) = f_t(p) \times \begin{cases} (1-\alpha)\eta + \alpha & p < b_t \\ 1 - \alpha - (1-\alpha)\eta & p \ge b_t \end{cases}$$

We set  $a_t$ ,  $b_t = p_t$ , i.e. the price at which the trade occurred was on the bid/ask. These probability distributions then need to be normalized.

When coding this model we create the sequence of vectors  $\{\mathbf{p}^t\}_{t\in\{0,1,...,T\}}$  such that each  $\mathbf{p}^t$  has  $|800\sigma|$  components with

$$p_i^t = P(\mu_t = m_0 - 4\sigma + i/100) = f_t(m_0 - 4\sigma + i/100)$$

This allows a discretization of  $\mu_t$  and its distribution  $F_t$  into the minimum increment (tick) of one cent. We denote the price  $m_0 - 4\sigma + i/100$  as  $v_i$  for convenience. The expectation  $\hat{\mu}_t$  can then be computed by

$$\hat{\mu}_t = \mathbb{E}F_t = \int_{-\infty}^{\infty} p f_t(p) dp \approx \sum_{i=0}^{\lfloor 800\sigma \rfloor} v_i p_i^t$$

$$\approx \sum_{i=0}^{\lfloor 800\sigma \rfloor} (m_0 - 4\sigma + i/100) f_t(m_0 - 4\sigma + i/100)$$