**Social Networks**

**Final Project**

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**Incentives to Cheat on Economic Announcements**

***1. Hypothesis:***

There is a large incentive to “cheat” around financial announcements such as non-farm payrolls, fed minutes, central bank policy, natural resource data, etc. The individuals familiar with this information have an incentive to leak the information to “financial agents” who can then act on this data before before its publicly announced. We can verify this hypothesis fast and cheaply by measuring the gains that a “financial agent” could make by prepositioning in the securities most impacted by the announcements. If there is no measurable difference between an agent who his prior knowledge of an event relative to an agent who could acquire then we can refute the hypothesis. If we can show there are gains to be made from acting on the knowledge beforehand then we can attempt to identify the scope of products and timings of optimal prepositioning.

***2. Background:***

While most insider trading headlines and enforcement activity center around stock trading, the insider market for macro trading/investing goes relatively unnoticed. Macro investing centers trading and investing in financial products that are impacted by macro economic events. These will include Interest Rate Futures, Foreign Exchange, Equity Futures, and Commodity Futures(Oil, Gold, Natural Gas, Crops, LiveStock, etc). Investors can tap into Futures Markets anywhere in the world to easily place bets on anticipated directional changes in any of the before mentioned contract types.

The largest price movements are typically centered around Central Bank Policy Updates and US Non-Farm Payrolls. Non-Farm Payrolls are released the first friday of every month at exactly 8:30 am (EST) and will provide the change in US Employment for the prior month. It is viewed as the most important number of economic outlook. Later in the month the Federal Reserve releases monthly FOMC minutes which outline current Federal Reserve Monetary Policy. FOMC minutes are released at exactly 2:00pm on one Wednesday of the Month.

There has long been suspicions that the important economic numbers and Fed Policy minutes are being leaked. Helping to fuel the suspicions is the cottage industry of economic forecasting firms run by academic economists. There are strong ties between academic economists and current Federal Reserve Officials, due the the large number of Economists who rotate between academia and the Fed. Questions have been raised as to whether trading firms paying for the forecasting services are actually paying for a forecast with large uncertainty or are they paying for access to economists who have strong ties the the Federal Reserve.

***3. Data:***

For analysis we used publicly available events calendar, Futures Tick Trade data from the CME and Twitter for President Trump’s communications. Futures where chosen for the analysis because they trade around the clock for 5.5 days out of the week and are highly liquid. We focused this study on Equity Futures (ES-E-mini S&P 500) , Short Term Interest Rate Futures (TU-2year Future) and Long Term Interest Rate Futures(UB-Ultra Long Bond). The analysis covers 2014 until April of 2019. Tick data was chosen over lower frequency to ensure consistency in the results and ensure maximum coverage of events as many events (non-farm payrolls for example) occur outside normal hours of market operation. Note: In general futures Tick data is not freely available, but can be purchased. Tick data for commercial purposes can be extremely expensive (historical tick data costs can well exceed $100,00). Researchers can find firms willing to give tick level data.

***4. Testing the Hypothesis***

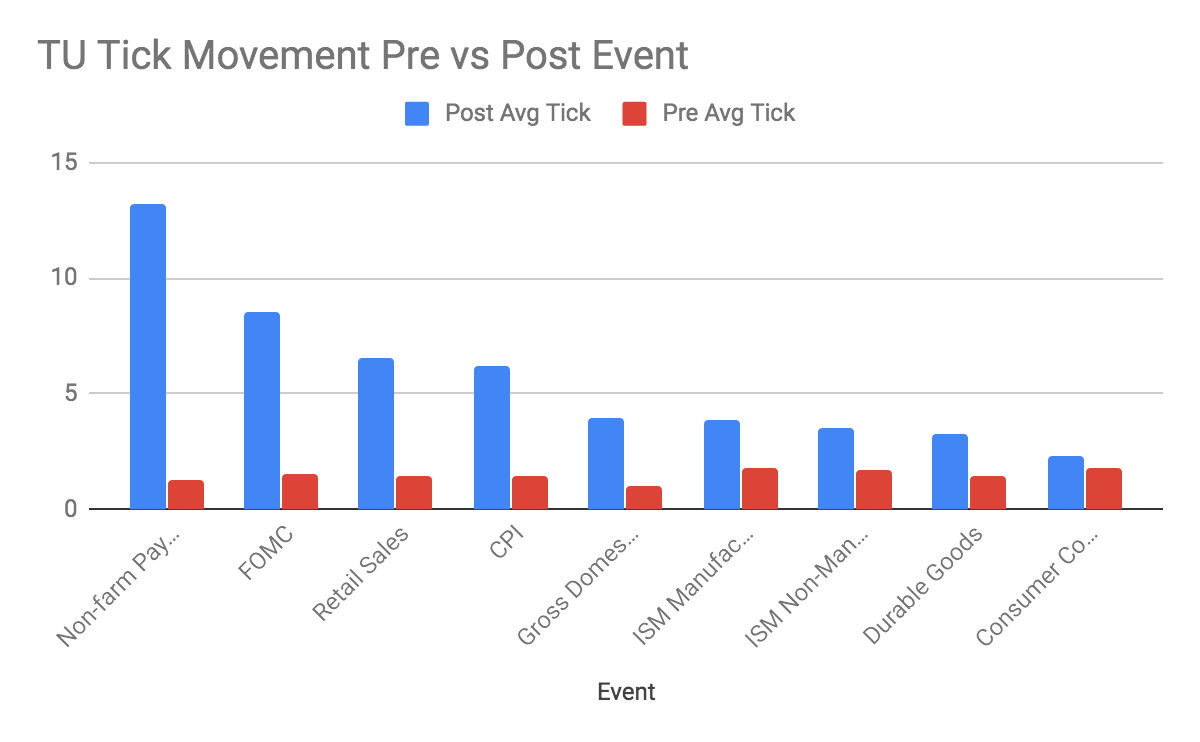
To test the hypothesis we want to examine the following questions:

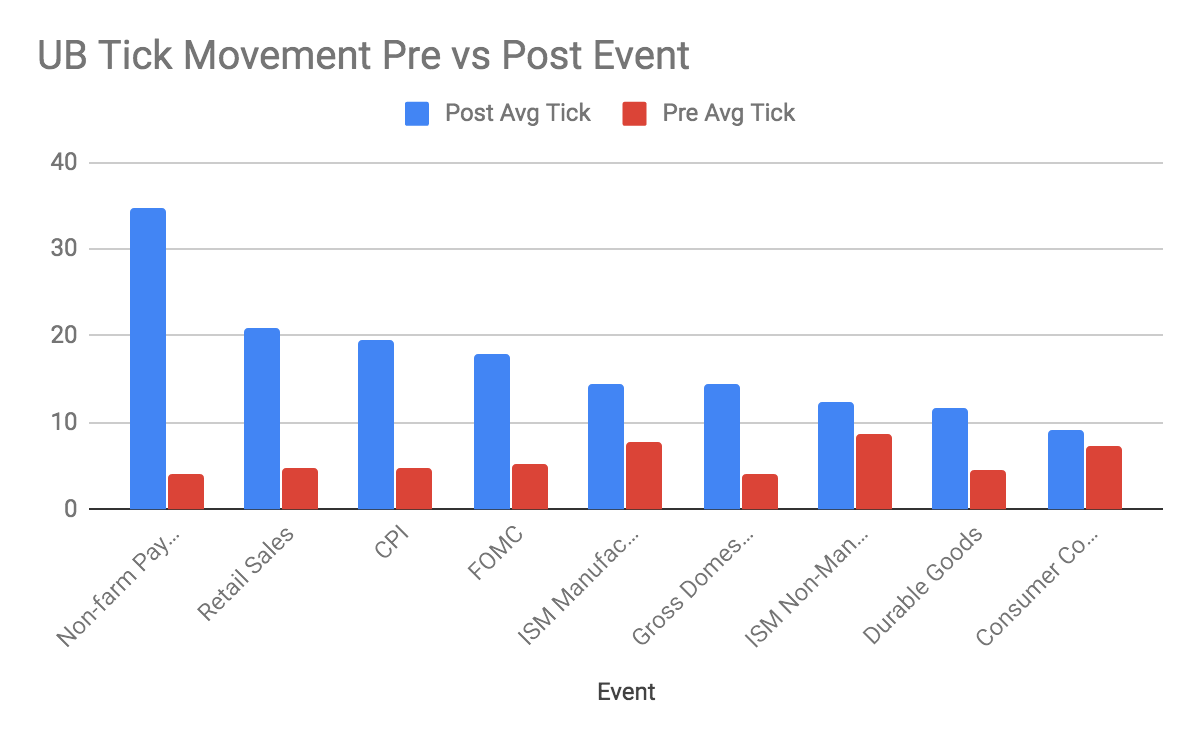
1. Do announcements present opportunity?
2. Can pre market movements help in predicting the direction of an event?
3. If we had the announcement data could we benefit from it?

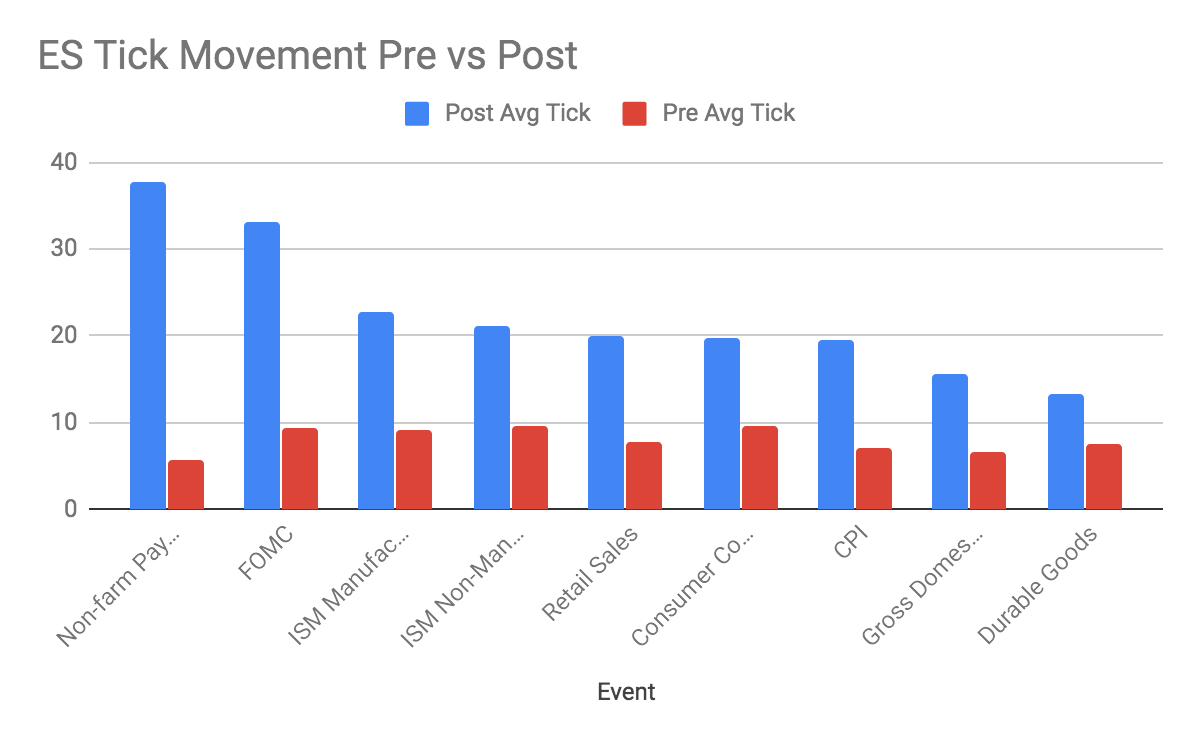
***4.1 Do announcements present opportunity?***

We will start with the premise there can be no crime without opportunity. To determine “opportunity” we will compare the magnitude of the market movement pre-event (for this study we chose a ten minute window and hour before each measure event) compared to post event ten minute window. Below are charts and tables comparing the pre-Event Windows and post-Event windows for nine of the most impactful economic announcements.

**The short answer: events present extremely outsized opportunities. It’s possible to make short term trades and gain outsized profits given the time period. The data supporting the opportunity size unambiguous.** The analysis data is presented in the tables and graphs below. The average tick movement across all events is approximately 3.5x larger than the average tick movement in prices across a similar time window without and event while the **trading volume is on average 6.5x larger and can be over 20x larger.** What's staggering is **maximum price movement in ticks can be over 29x the market average.** In the sample conducted over 99% of the tick market moves were greater than the non-event tick average. In general the most impactful events in this study are Non-Farm Payrolls with moves on average 8x the market and Federal Open Market Committee (FOMC)Rates Decision with moves over 5x the market. The largest single movement in the study on an ES Contract was due to (FOMC) announcement where the ES contract moved a staggering 199 ticks in 9 minutes, compared to a 7 tick move during the average window.







covers approximately 58 events from Jan 2104 to April 2019

**Event Tick Movement UB (Max,Mean), (No Event Mean) and Traded Volume Ratio**

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Symbol** | **Event** | **N** | **Event Max** | **Even Mean** | **No Event** | **Vol Ratio** |
| UB | Non-farm Payrolls | 58 | 80 | 35 | 4 | 16.4 |
| UB | Retail Sales | 58 | 46 | 21 | 5 | 8.5 |
| UB | CPI | 58 | 44 | 20 | 5 | 8.0 |
| UB | Gross Domestic Product | 46 | 29 | 14 | 4 | 5.9 |
| UB | FOMC | 54 | 45 | 18 | 5 | 3.6 |
| UB | Durable Goods | 56 | 29 | 12 | 5 | 4.2 |
| UB | ISM Manufacturing | 57 | 31 | 14 | 8 | 2.1 |
| UB | ISM Non-Manufacturing | 55 | 43 | 12 | 9 | 1.7 |
| UB | Consumer Confidence | 53 | 32 | 9 | 7 | 1.4 |

**Event Tick Movement TU (Max,Mean), (No Event Mean) and Traded Volume Ratio**

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Symbol** | **Event** | **N** | **Event Max** | **Mean** | **No Event** | **VolumeR** |
| TU | Non-farm Payrolls | 58 | 35 | 13 | 1 | 17.1 |
| TU | FOMC | 54 | 45 | 9 | 2 | 5.4 |
| TU | Retail Sales | 58 | 14 | 7 | 1 | 9.0 |
| TU | CPI | 58 | 13 | 6 | 1 | 8.5 |
| TU | Gross Domestic Product | 46 | 10 | 4 | 1 | 6.7 |
| TU | Durable Goods | 56 | 8 | 3 | 1 | 4.4 |
| TU | ISM Manufacturing | 57 | 10 | 4 | 2 | 2.7 |
| TU | ISM Non-Manufacturing | 55 | 12 | 3 | 2 | 2.1 |
| TU | Consumer Confidence | 53 | 12 | 2 | 2 | 1.8 |

**Event Tick Movement ES (Max,Mean), (No Event Mean) and Traded Volume Ratio**

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Symbol** | **Event** | **N** | **Max** | **Mean** | **No Event** | **VolumeR** |
| ES | Non-farm Payrolls | 58 | 138 | 38 | 6 | 14.3 |
| ES | FOMC | 54 | 199 | 33 | 9 | 3.6 |
| ES | CPI | 58 | 194 | 20 | 7 | 4.5 |
| ES | Retail Sales | 58 | 194 | 20 | 8 | 5.0 |
| ES | ISM Manufacturing | 57 | 65 | 23 | 9 | 7.5 |
| ES | Gross Domestic Product | 46 | 49 | 16 | 7 | 4.1 |
| ES | ISM Non-Manufacturing | 55 | 67 | 21 | 10 | 6.6 |
| ES | Consumer Confidence | 53 | 68 | 20 | 10 | 5.9 |
| ES | Durable Goods | 56 | 61 | 13 | 7 | 2.6 |

***4.2 Can pre-market moves help in predicting the direction of the event?***

To test this premise we exame a few different models to determine how much (if at all) pre-event returns can help in predicting post event returns. We evaluated the prediction of post-event return using two models and condition on different criteria, all pre-event returns and pre-event returns that were 1 standard deviation greater than the mean pre-event move. The first model predicts the post event return using a standard linear regression, the second model is a classification model to just predict the post move direction (up or down). We choose support vector machine because it has stricter criteria than a simple logistic regression.

The reasoning on **“Does someone know something?”.**

Since not every economic event results in large post announcement market moves, we decided to check to see if larger than normal pre event market moves were any different in predicting post event moves. A priori in absence of any information, there would no reason to believe that there would be any predictive difference in the ability of higher premarket moves to be able to predict post event market moves.

**Findings**

We were surprised to find that when the pre-event market move is greater than 1 standard deviation from the mean pre-event market move, the predictive power of the pre-market move is significantly better compared to all market moves. **A speculator has much better odds of making a profit when the pre-event movement is larger than usual!** This was actually surprising to find. Also the ability to predict post market returns from pre-market returns using all data for each event was close to zero, while the ability to predict post even direction fared better. The results are summarized below.

**Models and Results for 4.2**

**Model 1: linear regression of post-event return on pre-event return**

**Model 2: support vector machine logistic regression of post-event direction using pre-event return**

**Columns for the tables**

I - Instrument

Event - the economic even**t**

pR:1s/as - the ratio of the mean pre-event Return > 1 std dev / all pre-event Returns

acc1sd- the accuracy of logistic svm in predicting post return direction conditioning on pre-event moves greater than 1 std dev from the mean pre-event move

Accall - accuracy of prevent returns predicting post event returns

N1sd - number of data samples in pre-event moves greater than 1 std dev

Nall - number of all pre-event moves

R21sd - the R-squared of pre-event return > 1 std predicting post event return

R2all - the R-squared of prevent return predicting post event return

**Predicting ES post return and post move direction, using all data and a one standard deviation move.**

|  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- |
| **I** | **EVENT** | **pR:1s/as** | **acc1sd** | **accall** | **n1sd** | **nall** | **r21sd** | **r2all** |
| ES | CPI | 3.8 | 0.75 | 0.57 | 4 | 58 | 0.05 | 0.00 |
| ES | Consumer Confidence | 0.4 | 0.63 | 0.51 | 8 | 53 | 0.17 | 0.10 |
| ES | Durable Goods | 9.7 | 0.71 | 0.66 | 7 | 56 | 0.06 | 0.01 |
| ES | FOMC | 5.9 | 0.75 | 0.65 | 4 | 54 | 0.62 | 0.07 |
| ES | Gross Domestic Product | 3.4 | 0.78 | 0.57 | 9 | 46 | 0.03 | 0.01 |
| ES | ISM Manufacturing | 1.7 | 0.56 | 0.56 | 9 | 57 | 0.02 | 0.02 |
| ES | ISM Non-Manufacturing | 0.6 | 0.67 | 0.71 | 6 | 55 | 0.05 | 0.00 |
| ES | Non-farm Payrolls | 2.5 | 0.60 | 0.50 | 5 | 58 | 0.47 | 0.00 |

**Predicting TU post return and post move direction, using all data and a one standard deviation move.**

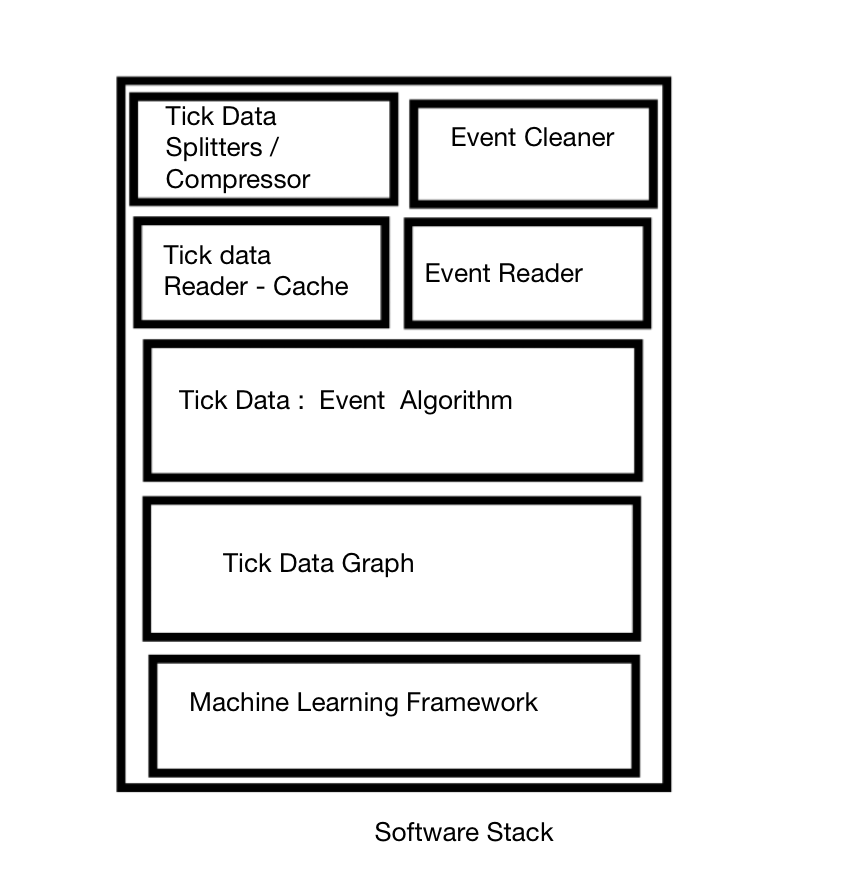
|  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- |
| **I** | **EVENT** | **pR:1s/as** | **acc1sd** | **accall** | **n1sd** | **nall** | **r21sd** | **r2all** |
| TU | CPI | 2.3 | 0.70 | 0.60 | 10 | 58 | 0.01 | 0.00 |
| TU | Consumer Confidence | 263.3 | 0.50 | 0.70 | 2 | 53 | 1.00 | 0.00 |
| TU | Durable Goods | 4.1 | 0.60 | 0.64 | 10 | 56 | 0.03 | 0.00 |
| TU | FOMC | 2.4 | 0.83 | 0.54 | 6 | 54 | 0.25 | 0.01 |
| TU | Gross Domestic Product | 42.8 | 0.57 | 0.78 | 7 | 46 | 0.01 | 0.01 |
| TU | ISM Manufacturing | 1.8 | 0.67 | 0.61 | 9 | 57 | 0.35 | 0.04 |
| TU | ISM Non-Manufacturing | 0.8 | 0.67 | 0.73 | 6 | 55 | 0.00 | 0.01 |
| TU | Non-farm Payrolls | 4.1 | 0.71 | 0.50 | 7 | 58 | 0.64 | 0.08 |
| TU | Retail Sales | 2.4 | 0.82 | 0.57 | 11 | 58 | 0.02 | 0.04 |

**Predicting TU post return and post move direction, using all data and a one standard deviation move.**

|  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- |
| **I** | **EVENT** | **pR:1s/as** | **acc1sd** | **accall** | **n1sd** | **nall** | **r21sd** | **r2all** |
| UB | CPI | 21.1 | 0.75 | 0.50 | 8 | 58 | 0.26 | 0.02 |
| UB | Consumer Confidence | 21.7 | 0.75 | 0.62 | 4 | 53 | 0.13 | 0.01 |
| UB | Durable Goods | 3.5 | 0.63 | 0.55 | 8 | 56 | 0.25 | 0.00 |
| UB | FOMC | 0.9 | 0.67 | 0.52 | 9 | 54 | 0.03 | 0.02 |
| UB | Gross Domestic Product | 0.0 | 0.57 | 0.52 | 7 | 46 | 0.03 | 0.00 |
| UB | ISM Manufacturing | 2.1 | 0.70 | 0.54 | 10 | 57 | 0.07 | 0.00 |
| UB | ISM Non-Manufacturing | 6.5 | 0.67 | 0.60 | 6 | 55 | 0.16 | 0.00 |
| UB | Non-farm Payrolls | 4.6 | 0.78 | 0.62 | 9 | 58 | 0.10 | 0.02 |
| UB | Retail Sales | 5.2 | 0.80 | 0.53 | 10 | 58 | 0.01 | 0.00 |

***6. The Software Stack***

The diagram below represents the software stack developed for this analysis which could turn into a product to lease to middle-tier financial firms. Many firms we have encountered do not have any in house capability to split, compresee, cache and then deliver tick data to a client program for analysis. The software takes in several GB of raw trade tick data from the CME, splits and compresses it so it can be quickly looked up via data time in a client analysis tool via the cache reader or used as input to simple machine learning framework developed for this project.

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***7. A Viable Product***

The software developed for this project has the potential to be licensed to financial institutions for analysis of any type of event. **In addition the software developed in this project can be turned into a website where the public can go to look at historical market behaviors when events occur.**

***8. Applications to other markets***

While we focused on US Futures Markets and Macroeconomic Events, the software stack developed can be applied to any exchange traded markets around the world. For instance we could use the methodology on earnings announcements on US Equities, international central banks announcements on their impact currency valuation and general foreign economic reporting and its impact on local equities and fixed income markets.

***9. Conclusion***

We have shown that economic announcements present outsized opportunity in terms of market volatility and trading volume compared to non-event market periods. We have show that higher than average pre-event market moves are better predictors of the post event price direction and returns compared to the average pre-event return. This lends some credence to the adage that “someone knows something.” In addition we have created the genesis of a product that can be used to analyze historical events impact on the markets.