

# Project HighLO and ROOT

May 16<sup>th</sup>  
PPP Meeting

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# Project HighLO – Key team

Collaboration: WUR + UM + CERN + CORMEC

HighLO = **H**igh Energy Physics Tools in **L**imit **O**rder Book Analysis

WUR/UM/CORMEC		CERN
Joost Pennings (WUR/UM)	Marjolein Verhulst (WU/WEcR)	Axel Naumann
Koos Gardebroek (WUR)	Philippe Debie (WU/WEcR)	Lorenzo Moneta
Bedir Tekinerdogan (WUR)	Tarek AISkaif (WUR)	Jonas Rembser
Cagatay Catal (WUR)		Danilo Piparo
Andres Trujillo-Barrera (CORMEC)		Han Dols

# Project HighLO – Background info

Current finance research lacks the tools

- Huge datasets
- Everything is statistically significant

HighLO

- Adapt more capable software tools (ROOT)
- A new perspective on how to analyze data
- Search and detect market manipulation

# Project HighLO – Background info

**Limit**

**Order**

**Book**

Side	Price	Volume
Ask Side (Selling)	15	$5 + 3 + 2 = 10$
	14	$4 + 2 = 6$
	13	$3 + 1 = 4$
Bid Side (Buying)	12	$2 = 2$
	11	$2 + 5 = 7$
	10	$5 = 5$

# Project HighLO – Background info

Financial market is governed by supply and demand

- Limit order book = Summary of all limit orders
- More sell limit orders → Price goes down
- More buy limit orders → Price goes up

Spoofing = Placing and cancelling limit orders to push the market  
(with no intention of execution)

# Project HighLO – Research goal

Detect spoofing in the commodity futures market

1. Describe how spoofing works
2. Detect spoofing
3. Help regulators and lawmakers

Spoofing = Placing and cancelling limit orders to push the market  
(with no intention of execution)

# Int. Expert Group on Market Surveillance (IMS)

## The Netherlands:

- Exchange: Euronext
- Regulator: Authority for Consumers and Markets
- Regulator: Authority Financial Markets

## UK:

- Exchange: ICE Futures Europe

## Germany:

- Exchange: Deutsche Börse (FSE, Eurex)
- Exchange: EEX

## Switzerland:

- Regulator: FINMA
- Exchange: SIX Group

## Italy:

- Regulator: CONSOB

## EU:

- Regulator: ACER
- Regulator: ESMA

## USA:

- Exchange: CME Group
- Regulator: CFTC

# Into the Microseconds

Paper submitted to Management Science



# Research Questions

1. How to measure relationships on the microsecond level?
  2. Can we measure which market is leading?  
And which market is following?
- This paper is not about market manipulation

# Data

	<b>Corn Futures</b>	<b>Wheat Futures</b>	<b>Soybean Futures</b>
CME Globex code	ZC	ZW	ZS
Data range	2019/07/01 – 2020/07/01		
Number of messages	758 Million	723 Million	1605 Million
Raw data	± 100 GB	± 100 GB	± 200 GB
In ROOT	5.3 GB	4.9 GB	9.9 GB

# Event-Based Impact Profile – The Concept

Idea:

Measure the changes in a time series  
at fixed intervals before or after an event

Example:

What is the average price change in the Wheat market  
50 ms after a price change in the Corn market

Trigger

Impact value

# Event-Based Impact Profile – The Construction

1. Collect a set of triggers

→ Price changes in the Corn market (800 k)

2. For each trigger, extract a time series

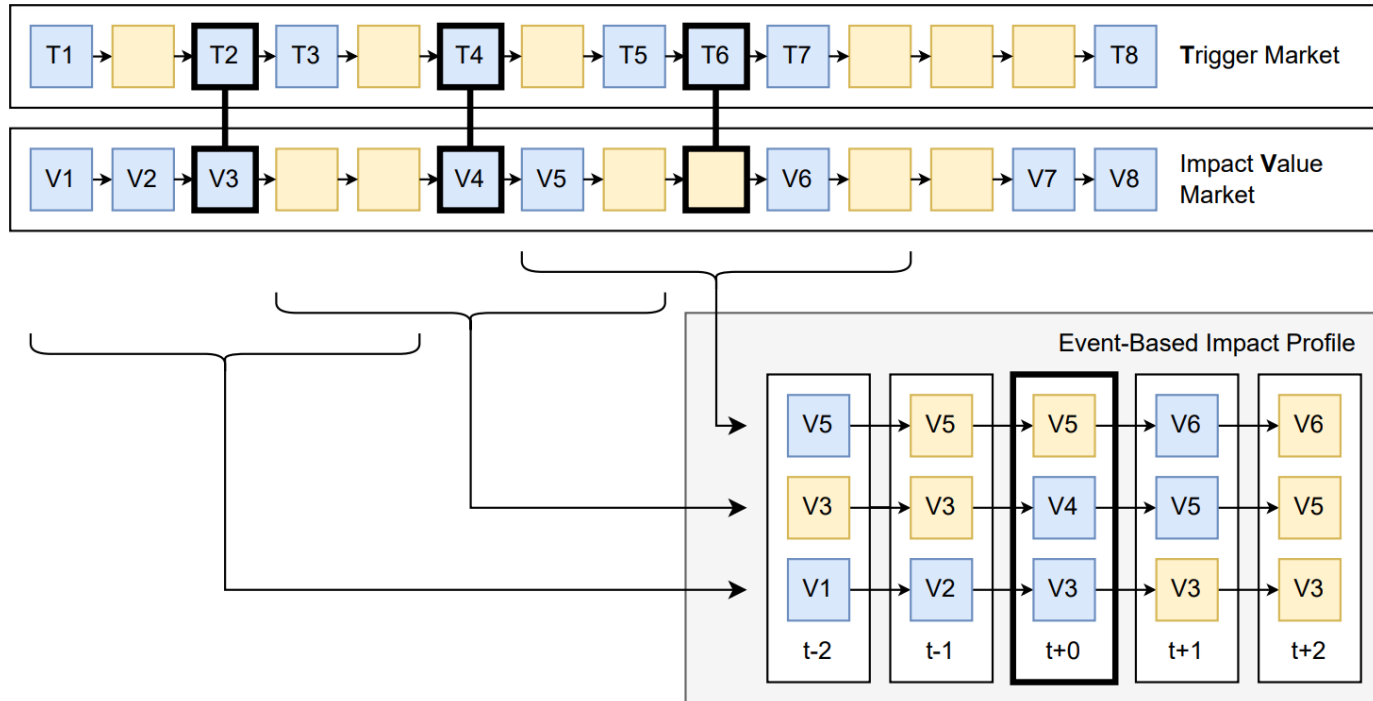
→ 200ms before to 200ms after the trigger

3. Overlap these time series (align the triggers)

4. For each time delay, build a distribution

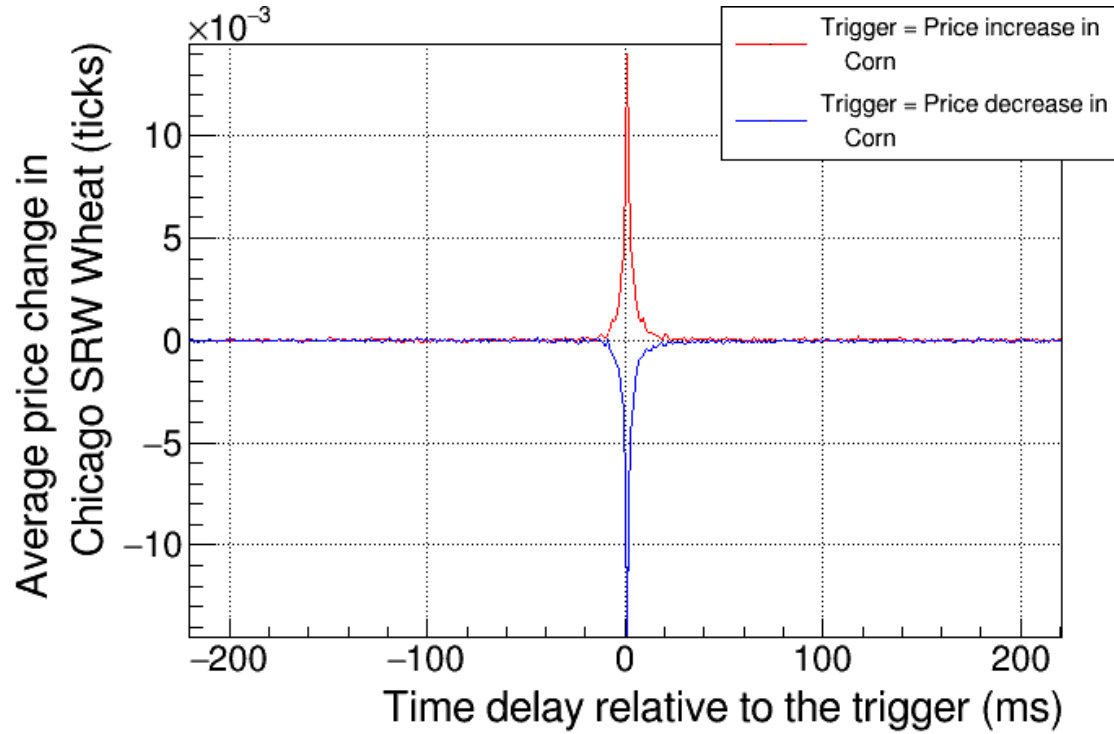
→ 401 distributions, sequence of distributions

# Event-Based Impact Profile – Diagram



Trigger:  
Impact:  
Resolution:

Price change in Corn Futures  
Price change in Wheat Futures  
1 ms (x-axis spanning 440 ms)



1 tick in Chicago SRW Wheat = 12.5 USD per contract (5000.0 BU)

Trigger:

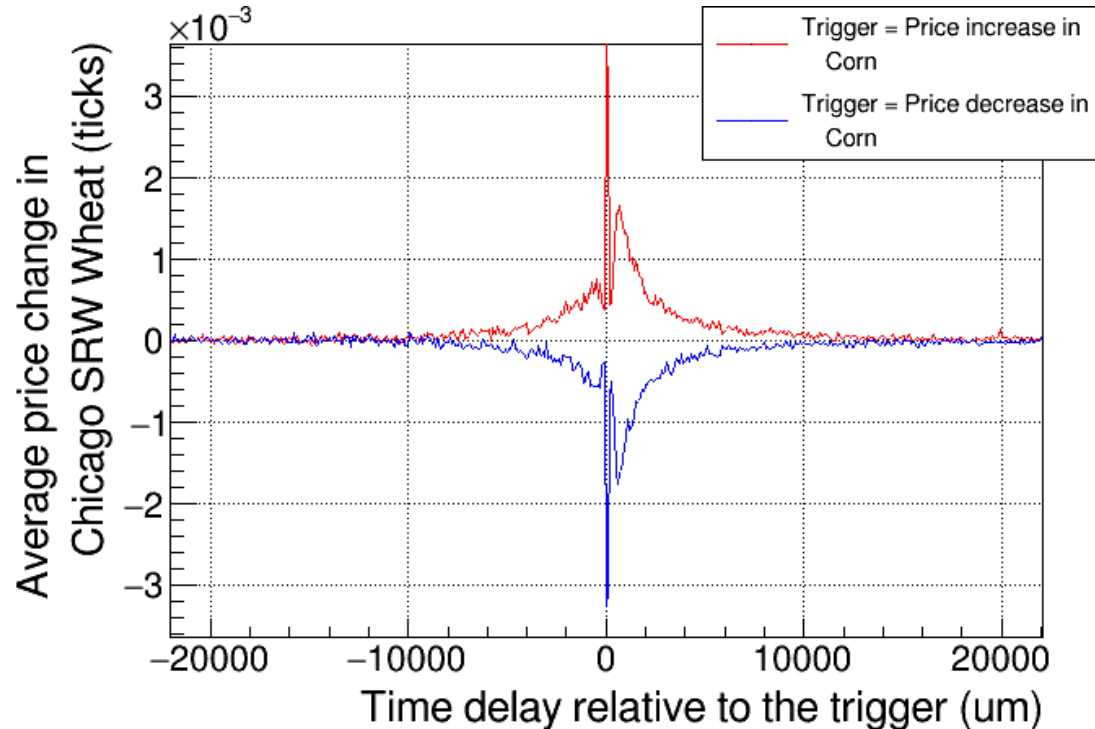
Price change in Corn Futures

Impact:

Price change in Wheat Futures

Resolution:

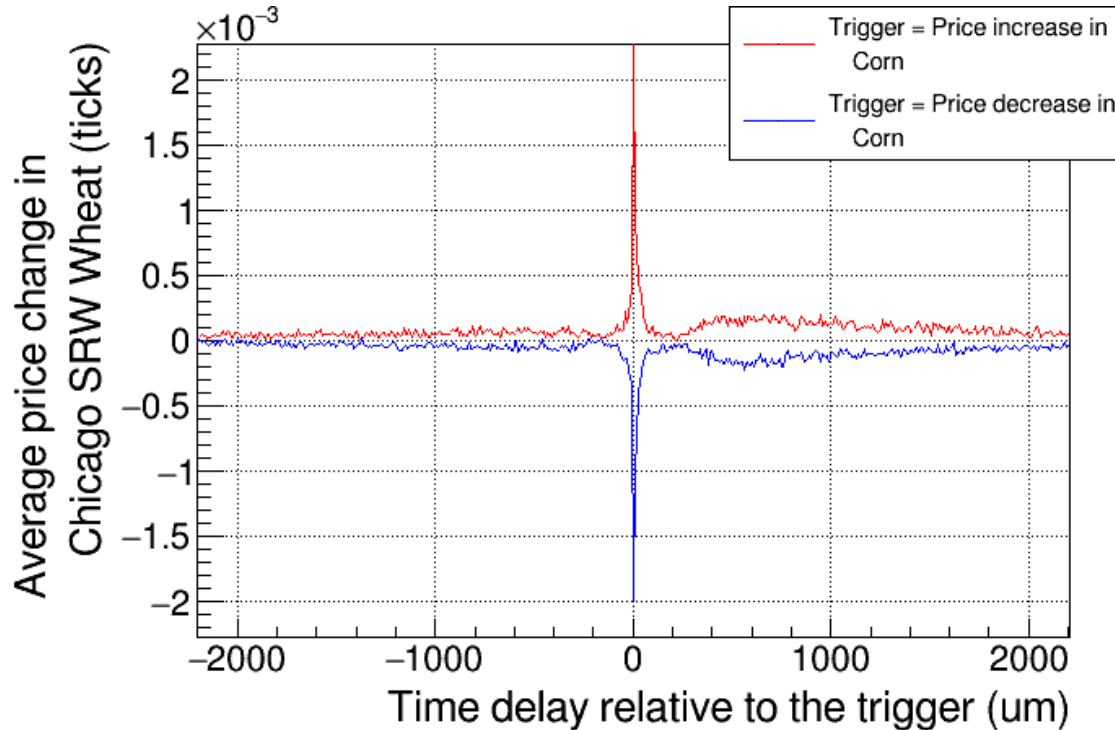
100  $\mu$ s (x-axis spanning 44 ms)



1 tick in Chicago SRW Wheat = 12.5 USD per contract (5000.0 BU)

Trigger:  
Impact:  
Resolution:

Price change in Corn Futures  
Price change in Wheat Futures  
10  $\mu$ s (x-axis spanning 4.4 ms)

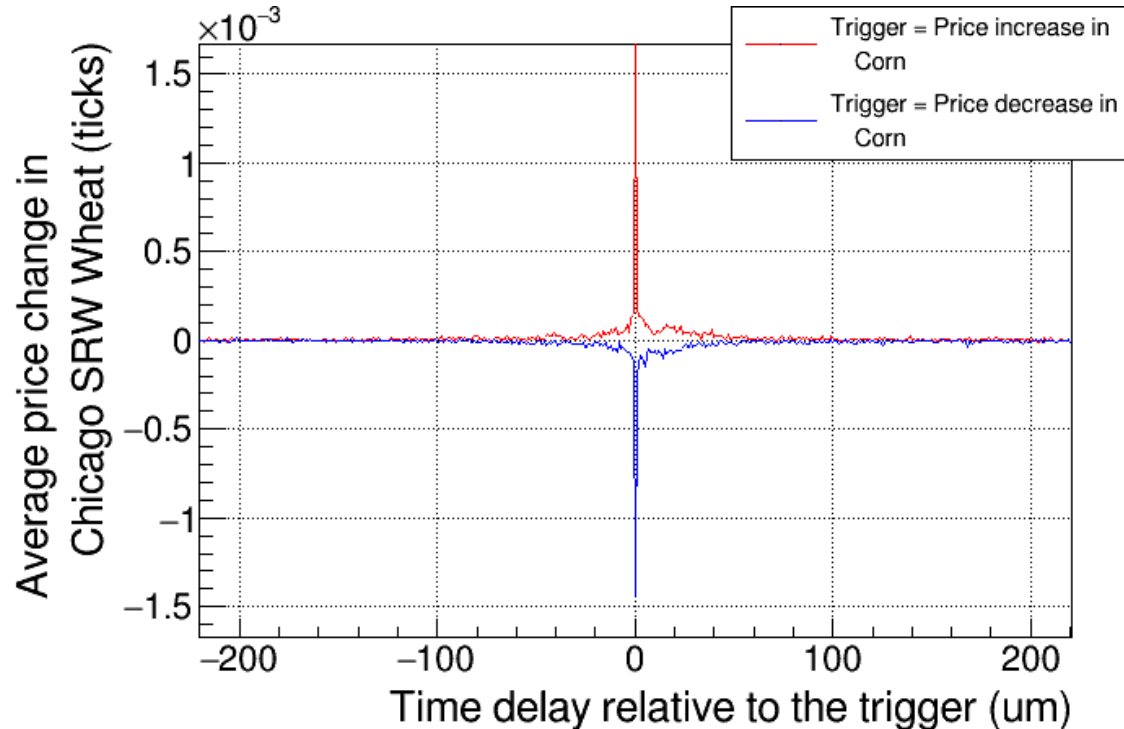


1 tick in Chicago SRW Wheat = 12.5 USD per contract (5000.0 BU)



Trigger:  
Impact:  
Resolution:

Price change in Corn Futures  
Price change in Wheat Futures  
1  $\mu\text{s}$  (x-axis spanning 0.44 ms)



1 tick in Chicago SRW Wheat = 12.5 USD per contract (5000.0 BU)

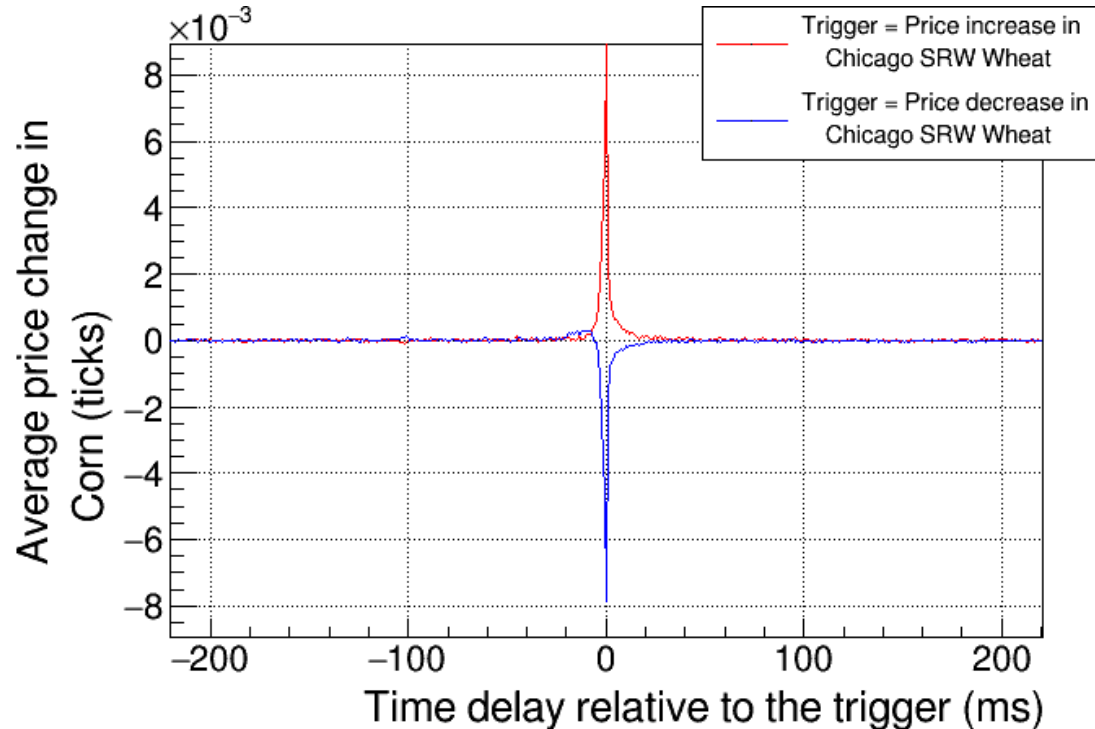
# Summary 1: Impact of Corn on Wheat

## Summary

1. Price increase occurs in 2 parts
  - a. Instantaneously
  - b. Starting 0.5 milliseconds after
2. Asymmetric, price increase in Wheat follows price increase in Corn

Next question: What if we inverse the trigger and impact contract?

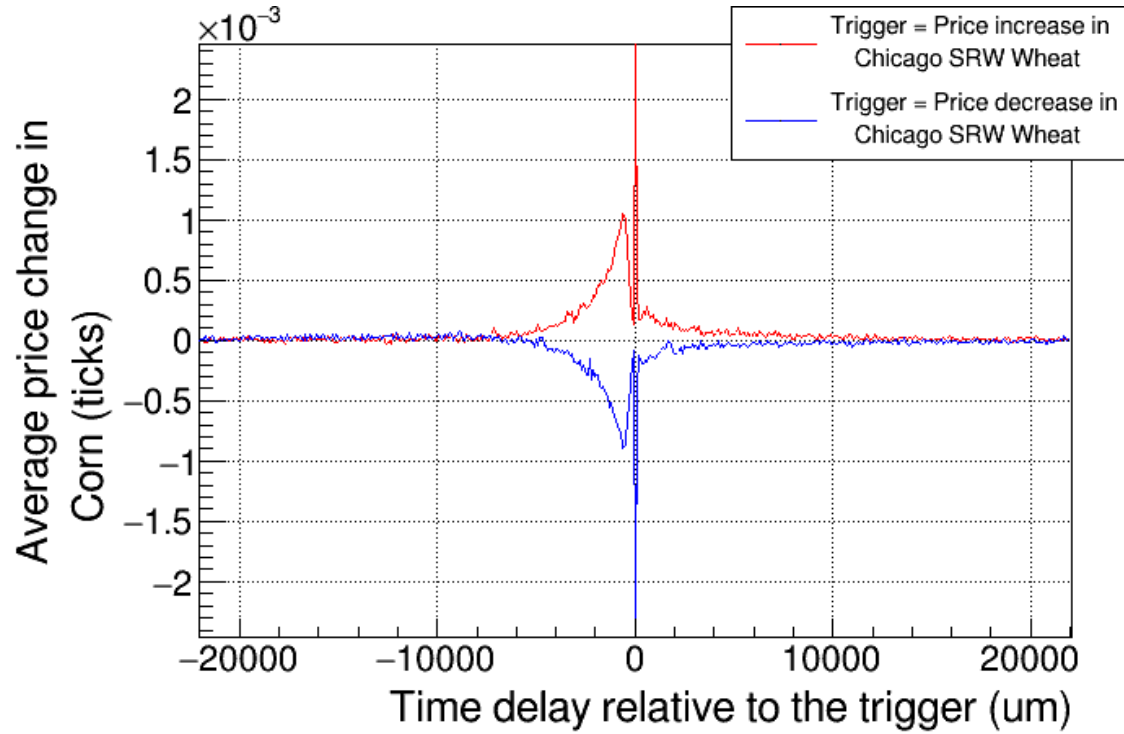
Trigger: Price change in Wheat Futures  
Impact: Price change in Corn Futures  
Resolution: 1 ms (x-axis spanning 440 ms)



1 tick in Corn = 12.5 USD per contract (5000.0 BU)

Trigger:  
Impact:  
Resolution:

Price change in Wheat Futures  
Price change in Corn Futures  
100  $\mu$ s (x-axis spanning 44 ms)



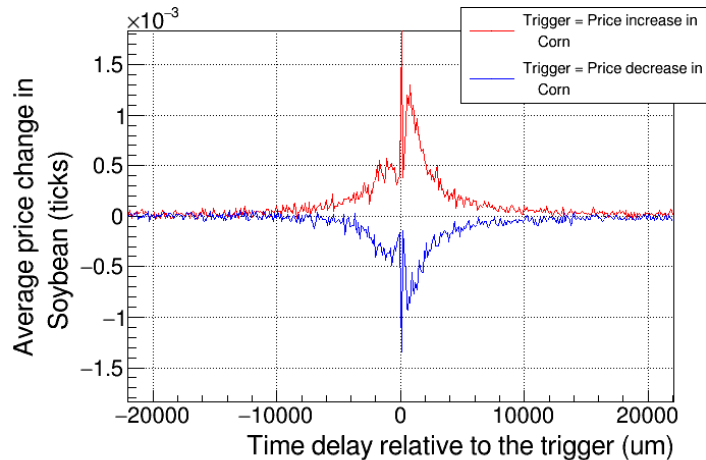
1 tick in Corn = 12.5 USD per contract (5000.0 BU)

# Summary 2: Impact of Wheat on Corn

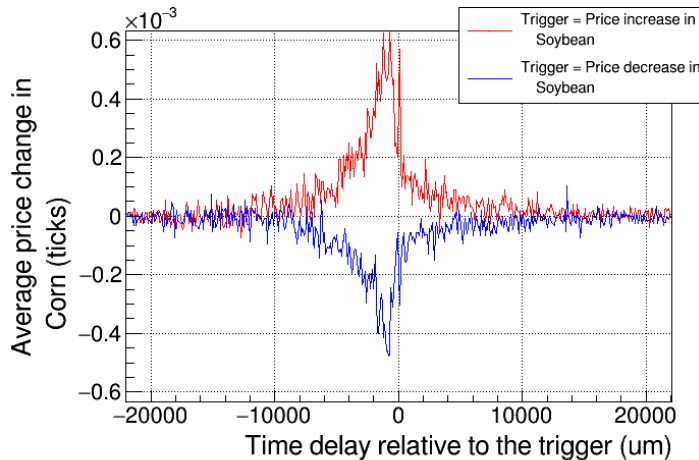
## Summary

1. Price increase occurs in 2 parts
2. Asymmetric, price increase in Wheat follows price increase in Corn

Next question: How are relations between other markets?

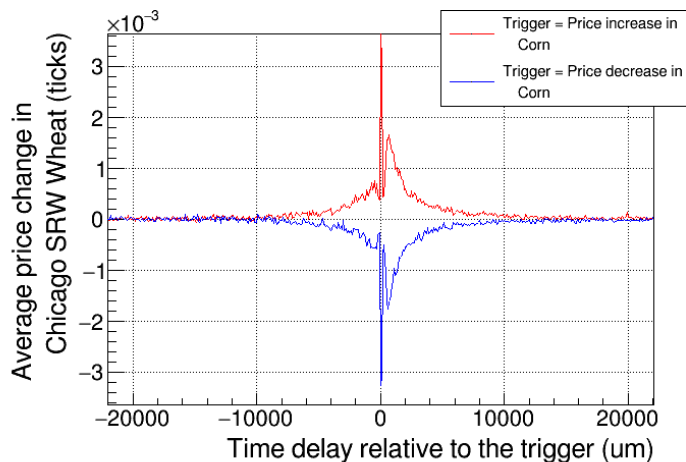


1 tick in Soybean = 12.5 USD per contract (5000.0 BU)

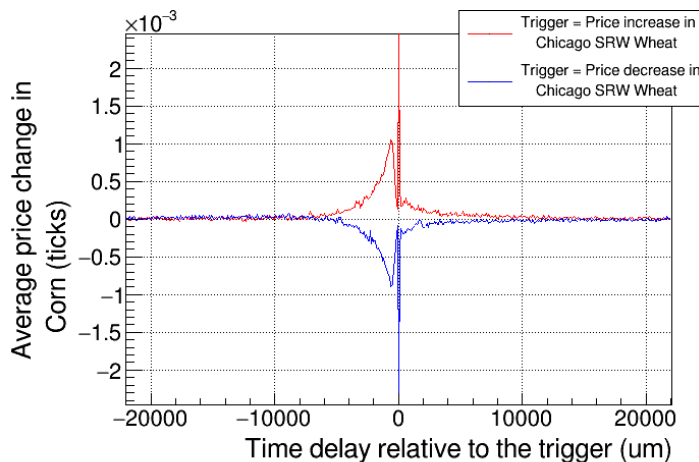


1 tick in Corn = 12.5 USD per contract (5000.0 BU)

Soybean follows Corn

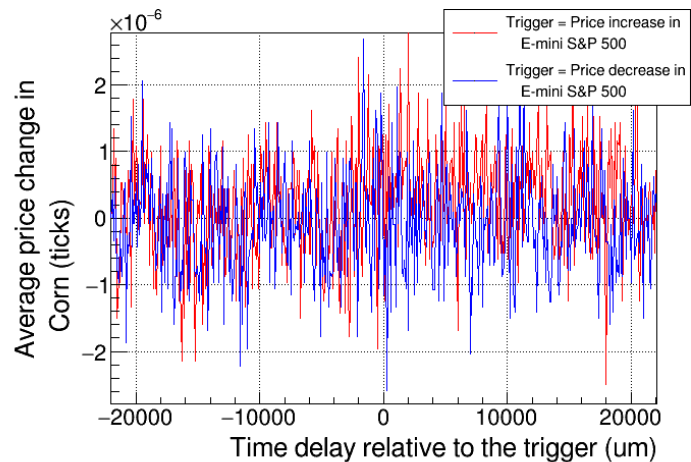


1 tick in Chicago SRW Wheat = 12.5 USD per contract (5000.0 BU)

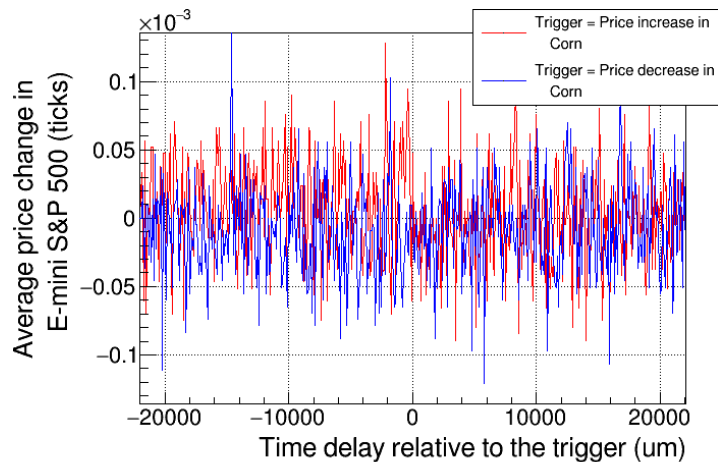


1 tick in Corn = 12.5 USD per contract (5000.0 BU)

Wheat follows Corn



1 tick in Corn = 12.5 USD per contract (5000.0 BU)



1 tick in E-mini S&P 500 = 12.5 USD per contract (50.0 IPNT)

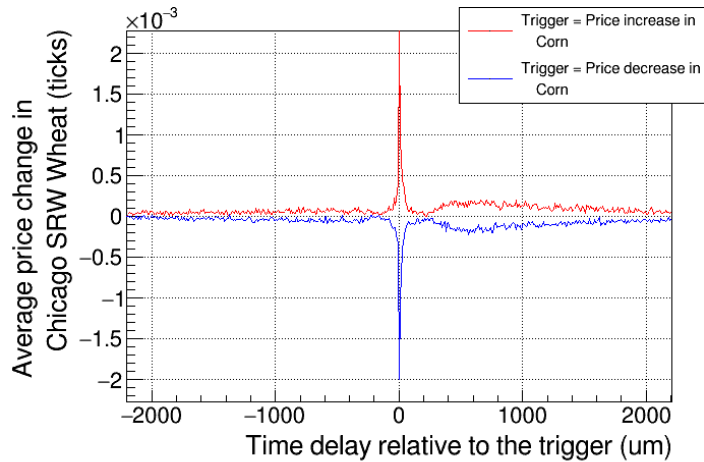
# Conclusion: Research question

Observations:

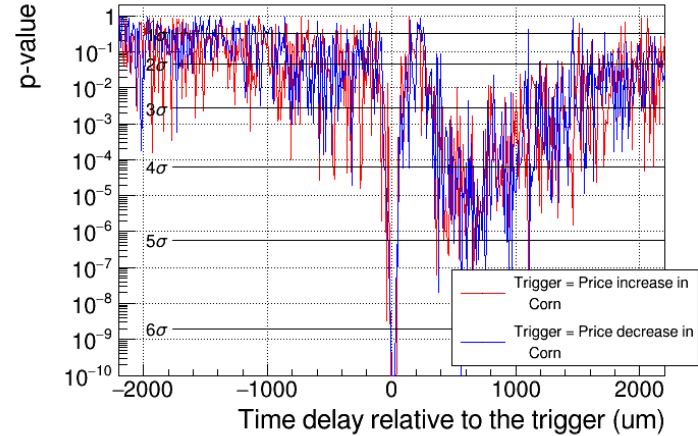
1. The market responds in 2 steps
2. **Price** changes in Wheat and Soybean follow **Price** changes in Corn



Trigger: Price change in Corn Futures  
Impact: Price change in Wheat Futures  
Resolution: 10  $\mu$ s (x-axis spanning 4.4 ms)



1 tick in Chicago SRW Wheat = 12.5 USD per contract (5000.0 BU)



# Conclusion: The methodology

Advantages compared to an impact analysis using VAR models

1. No data fitting or user-chosen modelling parameters
2. Measure past and future time correlations
3. Linear computational cost and trivial multi-threading
4. Not limited by the complexity of the model

→ Interdisciplinary collaboration led to new techniques

# How does HighLO contribute to ROOT?

Project 1: Interactive Dashboard to Explore High-Dimensional Histograms

# How does HighLO Contribute to ROOT?

Project 2: RDataframe with Time Series Support

# New Operations for RDataframe

Slides from PPP 118

## RDataframe operations

- Define using lead and lag (differentiation)
- Persistent data objects (integration and more)
- Resample a time series

Proof of concept: <https://github.com/philippe554/root>

# Lead and Lag

```
ROOT::RDataFrame rdf(50);

auto r = rdf
  .DefineSlotEntry("foo", [](unsigned int slot, ULong64_t entry){return static_cast<int>(entry);})
  .Define("bar", [](int foo){return foo * foo;}, {"foo"})
  .MovingCache<int, int>({"foo", "bar"})
  .Define("D", [](int bar1, int bar2){return bar2 - bar1;}, {"bar", "bar"}, {-1, 0})
  .Display({"foo", "bar", "D"});

r->Print();
```

```
+-----+-----+-----+-----+
| Row | foo | bar | D |
+-----+-----+-----+-----+
| 1   | 1   | 1   | 1 |
+-----+-----+-----+-----+
| 2   | 2   | 4   | 3 |
+-----+-----+-----+-----+
| 3   | 3   | 9   | 5 |
+-----+-----+-----+-----+
| 4   | 4   | 16  | 7 |
+-----+-----+-----+-----+
```

→ Note that it skipped the first entry

# Persistent Define

```
ROOT::RDataFrame rdf(10);

auto r = rdf
    .DefineSlotEntry("foo", [](unsigned int slot, ULong64_t entry){return static_cast<int>(entry);})
    .Define("D", [](){return gRandom->Exp(1);})
    .DefinePersistent("time", [(double& time, double D){time += D;}, {"D"}])
    .DefinePersistent("state", [(std::string& state, int foo){state = state + std::to_string(foo);}, {"foo"}])
    .Display({"foo", "D", "time", "state"});

r->Print();
```

```
+-----+-----+-----+-----+-----+
| Row | foo | D           | time           | state          |
+-----+-----+-----+-----+-----+
| 0   | 0   | 0.00025828445 | 0.00025828445 | "0"           |
+-----+-----+-----+-----+-----+
| 1   | 1   | 1.8145581    | 1.8148164    | "01"          |
+-----+-----+-----+-----+-----+
| 2   | 2   | 1.2636598    | 3.0784762    | "012"         |
+-----+-----+-----+-----+-----+
| 3   | 3   | 0.054243873  | 3.1327201    | "0123"        |
+-----+-----+-----+-----+-----+
| 4   | 4   | 1.4624994    | 4.5952195    | "01234"       |
+-----+-----+-----+-----+-----+
| 5   | 5   | 0.72366079   | 5.3188803    | "012345"      |
+-----+-----+-----+-----+-----+
```

# Resample a Time Series

Slides from PPP 118

```
ROOT::RDataFrame rdf(50);

auto r = rdf
  .DefineSlotEntry("foo", [](unsigned int slot, ULong64_t entry){return static_cast<int>(entry);})
  .Define("D", [](){return gRandom->Exp(1);})
  .DefinePersistent("time", [(double& time, double D){time += D;}, {"D"})
  .Resample<double, double, int>("time", 1, 5, 15, {"time", "foo"})
  .Display({"time", "foo"}, 10);

r->Print();
```



# Resample a Time Series

Slides from PPP 118

```
+-----+-----+-----+
| Row | time      | foo |
+-----+-----+-----+
| 0   | 5.000000 | 5   |
+-----+-----+-----+
| 1   | 6.000000 | 6   |
+-----+-----+-----+
| 2   | 7.000000 | 7   |
+-----+-----+-----+
| 3   | 8.000000 | 8   |
+-----+-----+-----+
| 4   | 9.000000 | 8   |
+-----+-----+-----+
| 5   | 10.000000| 9   |
+-----+-----+-----+
| 6   | 11.000000| 9   |
+-----+-----+-----+
| 7   | 12.000000| 9   |
+-----+-----+-----+
| 8   | 13.000000| 10  |
+-----+-----+-----+
| 9   | 14.000000| 11  |
+-----+-----+-----+
```

# Discussion