



## **UConn top 25 Public Research Institution**

92% secured post – graduate employment or service

2,249 undergrads, 329 grads, 2,578 total

70% are of diverse backgrounds, 52% are first gen, 47% are Pell grant eligible

## **UConn Stamford Computer Science program started in 2016**

Focus on Internships and Real-World Challenges

Synchrony Financial build a co-location lab on Campus hired 20+ students

Financial companies -Morgan Stanley, UBS, RBS, Mckinsey, AQR and Point 72.

# State Street Project Team



**Dr. Phil Bradford**

20+ Years Financial Services  
Blackrock, GE, State Street  
22+ Teaching  
Univ Alabama, Rutgers, UCONN

**John Iacovacci, MBA**

20+ Years Financial Services  
Citigroup, GE, State Street  
20+ Teaching  
Fordham , UCONN



**Haris Mahmood**  
Google DSC Lead  
UCONN Senior  
The Hartford



**Shalini Kuzhippat**  
UCONN Senior  
Computer Science



**Charlton Toussaint**  
UCONN Senior  
Computer Science



**Charlie Kuryluk**  
UConn Student  
Cyber Security



**Dedeep Singu**  
UCONN CS Major  
Math Minor  
intern Synchrony

**Rauni Machado-Filho**  
UCONN Senior

**David Keenan**  
UCONN Junior



# UConn Stamford Google Club


Google Developer Student Clubs for students interested in Google technologies.

Stamford Google Club - only club in CT - 170 students

## Speakers

GE, Bloomberg, Lockheed Martin, Google, Twitter, Deutsche Bank, State Street Bank.  
Cloud development platform creating real-world applications for financial sector.



 Google Developer Student Clubs  
University of Connecticut





## Susan Lasota

Chief Technology Officer  
STATE STREET GLOBAL ADVISORS

Thu. Oct. 13 | 5pm EST | Auditorium 1

- Learn how to advance your career in tech
- State Street manages over 3.4 trillion dollars in assets
- Potential job/internship opportunities at State Street!
- Free Pizza and Drinks



RSVP Today!

# Public Equities – Sentiment Project

- Build a library of key words and topics for sentiment model
- Build and develop sentiment model to analyze company conference call transcripts to identify shift in company fundamentals
- Develop sentiment score using sentiment model analysis
- Train sentiment model to avoid « hallucination”
- Incorporate sentiment score into Fundamental Equity team CQ Calibration dashboard (in Spotfire)

**Ability to monitor conference calls for covered companies and identify sentiment of categories:**

- |                    |              |
|--------------------|--------------|
| - Financial metric | - Macro      |
| - Sector trend     | - Regulation |



# Keywords / Topics

Inflation  
Interest rate  
Raw Materials inflation  
Volume  
Revenue  
Earnings  
Earnings per Share / EPS  
Share Repurchase  
Buy back  
Dividends  
Economy  
Operating Expenses  
Operating margins  
Pricing  
Gross margin  
Market share  
Employee attrition  
Inventory  
Fuel cost  
Wage increase  
Price increases  
Guidance  
Sales drivers  
Margin improvement

Revenue Growth  
Competitive Environment  
EBIT Margins  
Organic growth  
Book-to-bill  
Contribution margin  
Orders  
Restructuring  
Inventory destocking  
Backlogs  
Order rates  
Incremental margins  
Recurring revenue  
Aftermarket growth  
Labor shortage  
Capacity utilization  
Economic stimulus  
Renewables  
Megaprojects  
Reshoring  
Regulation  
Capex  
Lay offs  
Unemployment

Net Interest Income  
Mortgage Fees  
Reserves  
Investment Banking Fees  
Trading revenues  
Credit cost  
Delinquency  
Capital Requirements  
Loan-to-Value  
Interest coverage  
Defaults  
Commercial Real Estate  
Collateral value  
Net Interest Margin (NIM)  
Loan growth  
Cost/income ratio  
Provisions  
Non-performing loans  
Credit card balance  
Delinquencies  
Mortgage Refinancing  
Debt issuance  
IPO  
Deposit

# Sentiment Scoring

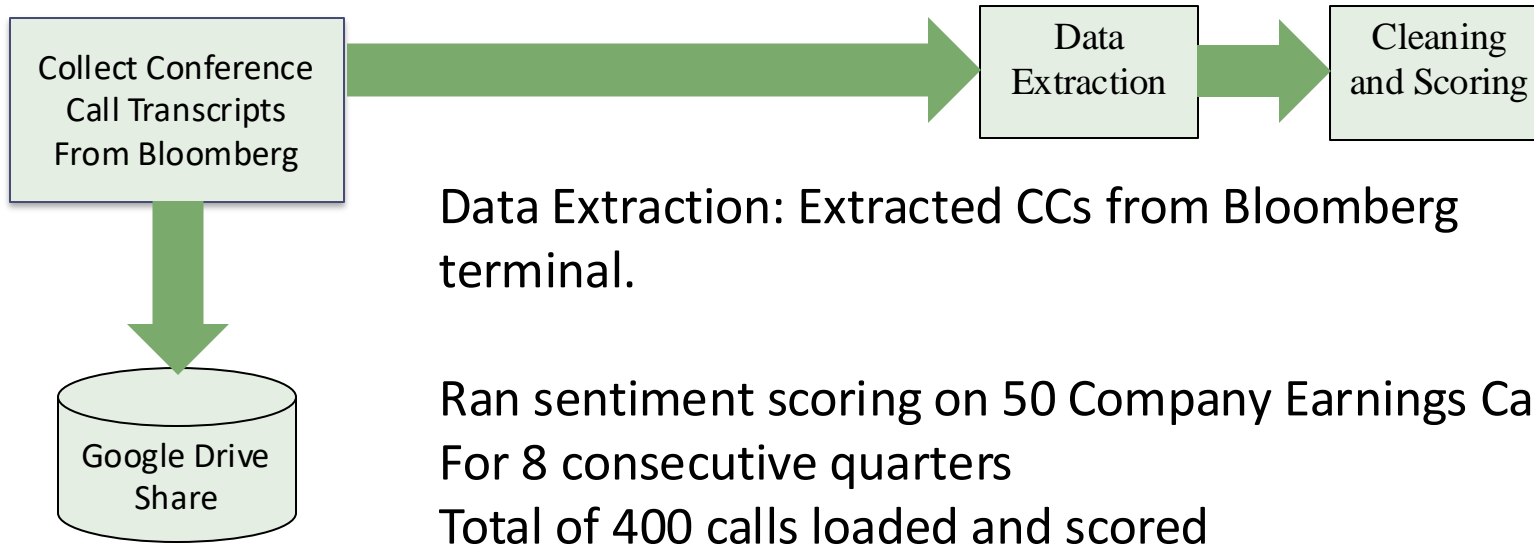
Conference calls are loading into the process and searched by keyword depending on sector. Paragraphs containing keywords are scored via sentiment model and weighted based on the importance of the word.

Sector	Categories	Weights (adjustable)
Industrials	Financial metric	Very important = 1.5
Banks	Macro	Important = 1
All	Sector trend	Less so important = 0.5
	Regulation	

Keywords/Topics	Sector	Key Word Category	Weights
Revenue	All	Financial metric - All	Very Important
Earnings per Share / EPS	All	Financial metric - All	Very Important
Share Repurchase	All	Financial metric - All	Important
Buy back	All	Financial metric - All	Very Important
Dividends	All	Financial metric - All	Important
Net Interest Income	Banks	Financial metric - Bank	Very Important
Economy	All	Macro	Less so important
Mortgage Fees	Banks	Financial metric - Bank	Important
Reserves	Banks	Financial metric - Bank	Very Important
Operating Expenses	All	Financial metric - All	Important
Credit cost	Banks	Financial metric - Bank	Important
Delinquency	Banks	Sector trend	Important

# Sentiment Loading Process

Conference call transcripts for a two-year period for each company



Documents stored by  
ticker, period, date of call.

e.g. WSP.TO\_Q123\_5-11-23





# Conference Call Sentiment score validation

Team performed manual score review which validated our automatic testing.

Individually scored Conference calls using generic Google sentiment scoring process results were in line with the BERT financial trained model we used.

Paragraph	Cloud NLP API Score	Model Score	Model Score weighted
Let me now address some addit	0.558	0.613	0.829
ber two in leverage finance with	0.837	0.945	0.999
And so those investments across	0.895	0.786	0.915
I'd also like to caution you that v	-0.023	-0.044	0.483



Positive



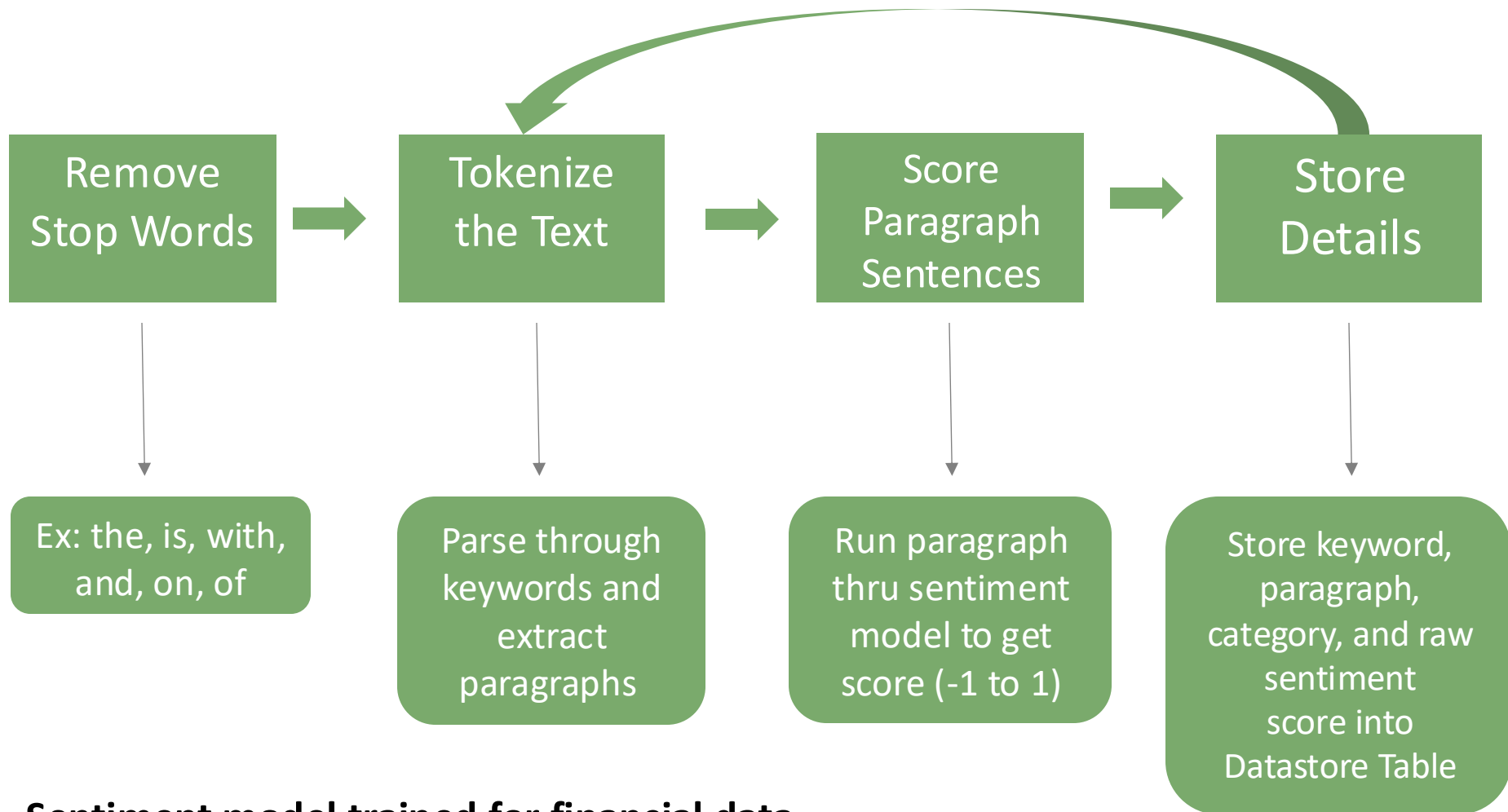
Negative



Neutral



# Sentiment Scoring Process



**Sentiment model trained for financial data**

# Sample of Stored Details

**Call Date:** 10/14/2021

**Ticker:** C US

**Category:** Financial metric

**Period:** Q42021

**Document Type:** CC

**Score:** 0.265

**Keyword:** Dividends

**Sector:** All

**Paragraph:** Let me briefly touch on our US retail strategy. Now, our digital strategy and investments are starting to pay dividends. Digital deposits stand at \$19 billion, having grown 26% in the last year. And we see more than two-thirds of our digital deposit balances coming from customers outside of our branch footprint, with about half of those deposit from our card holders who did not previously have a retail relationship with us.

**Weighted Sentiment:** 0.397

# Sample of Sentiment Summary:

**Ticker: C US**

**Period: Q42021**

Summarize  
Sentiment



**Financial metric - Average: 0.014**

**Financial metric - Count: 32**

**Financial metric - Weighted Average: 0.035**

**Macro Average: 0.402**

**Macro Count: 3**

**Macro Weighted Average: 0.361**

**Regulation Average: 0**

**Regulation Count: 0**

**Regulation Weighted Average: 0**

**Sector trend Average: 0.383**

**Sector trend Count: 8**

**Sector trend Weighted Average: 0.321**

**Total Average: 0.110**

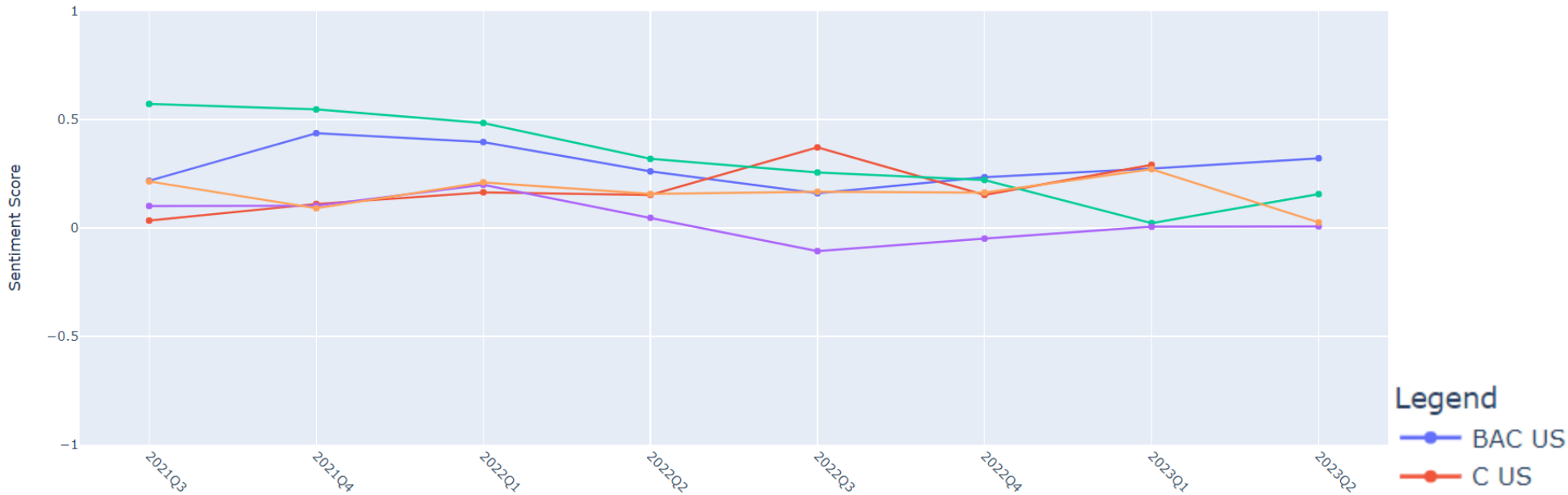
**Weighted Average: 0.111**

Aggregate  
Sentiment  
Details into  
Summary Table

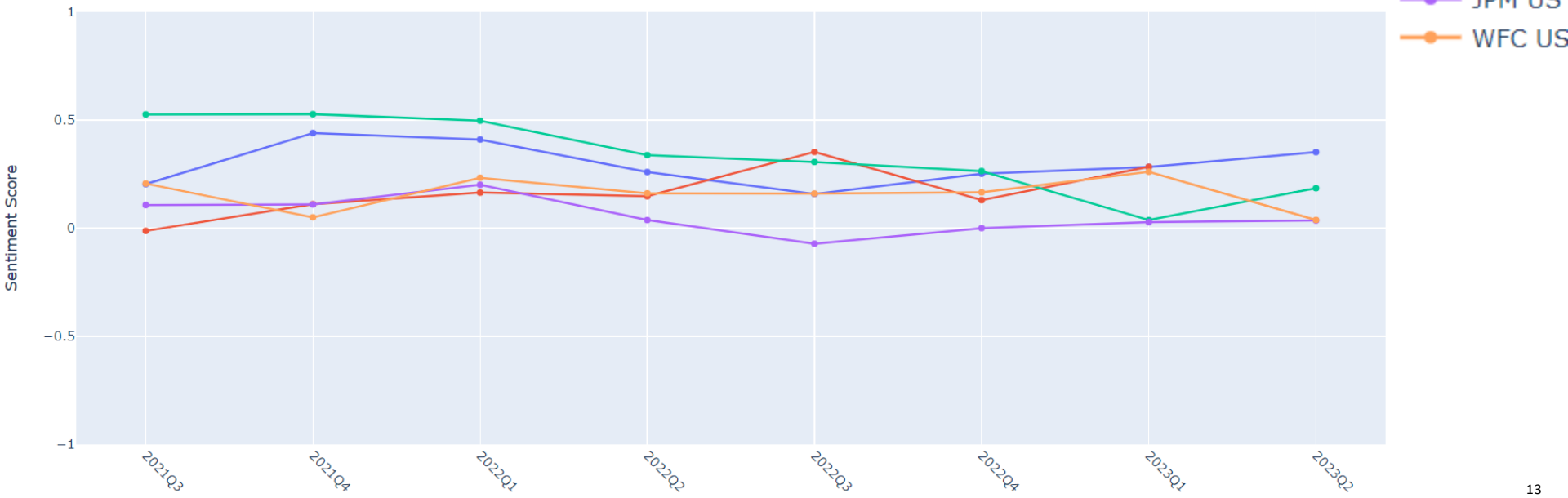


# Comparison Graphs (Banks)

## Raw score

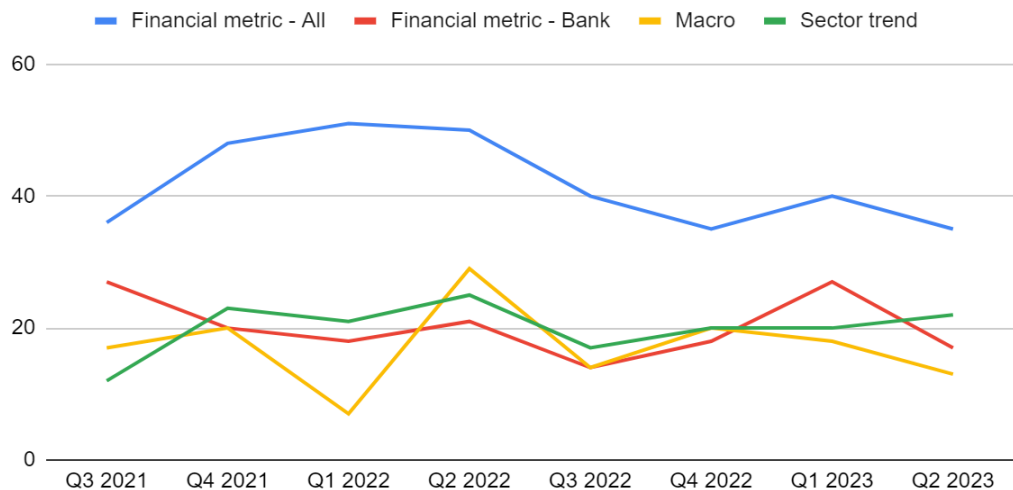


## Weighted score

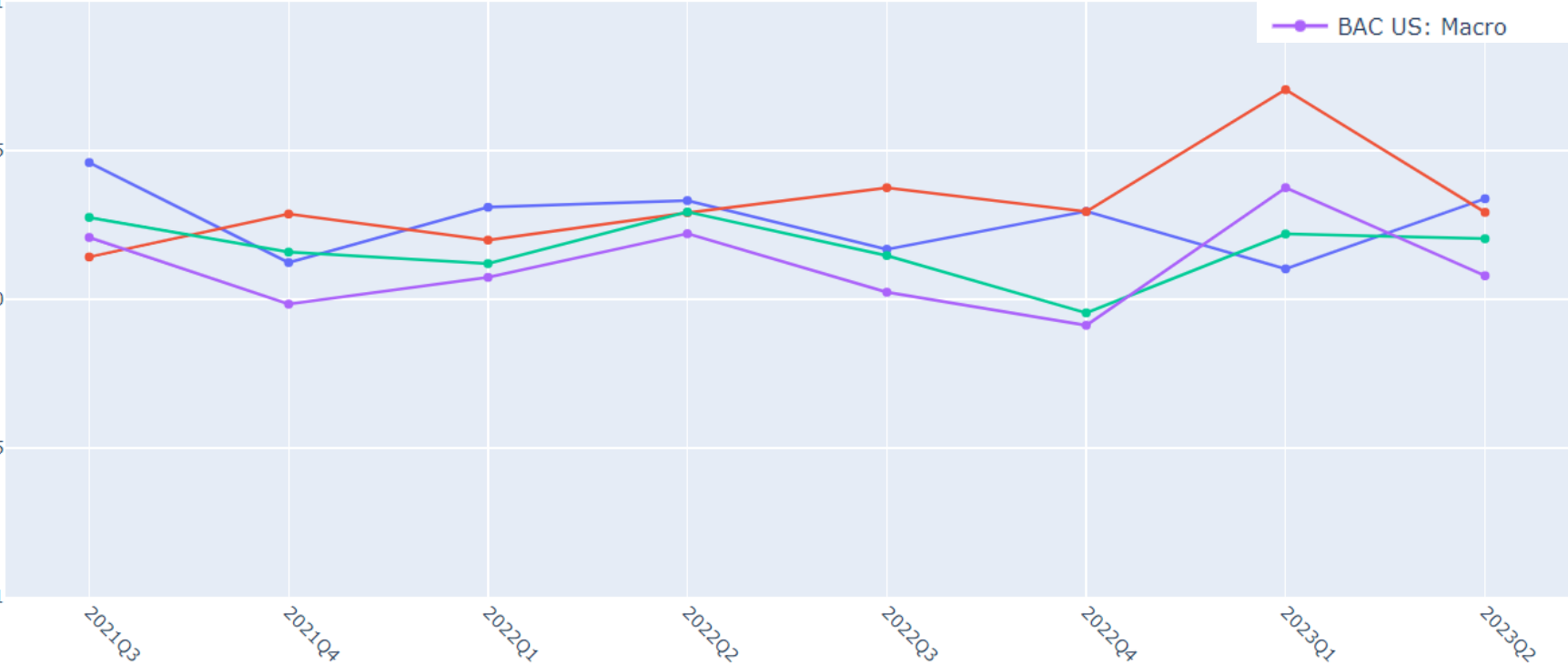


# Keyword Occurrences – BAC US

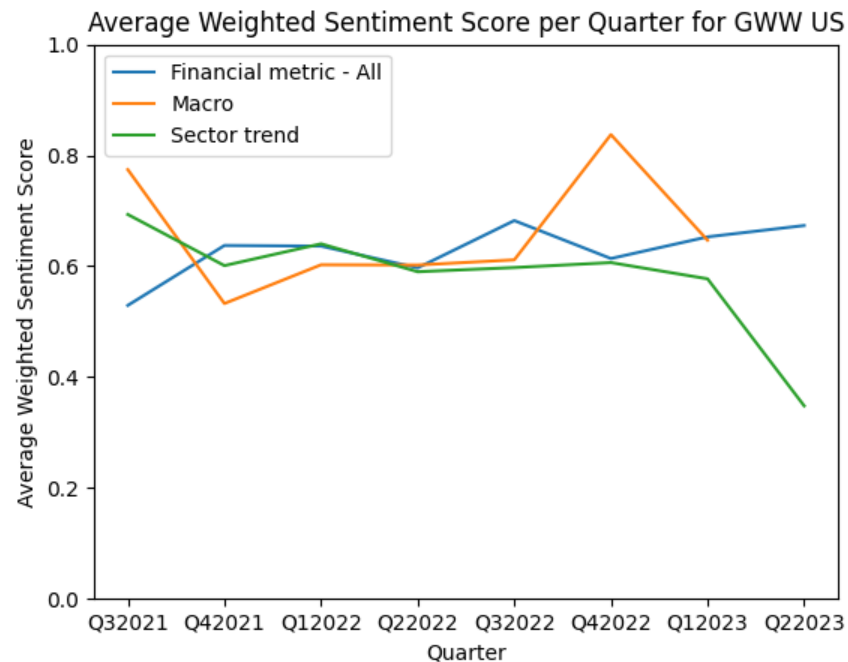
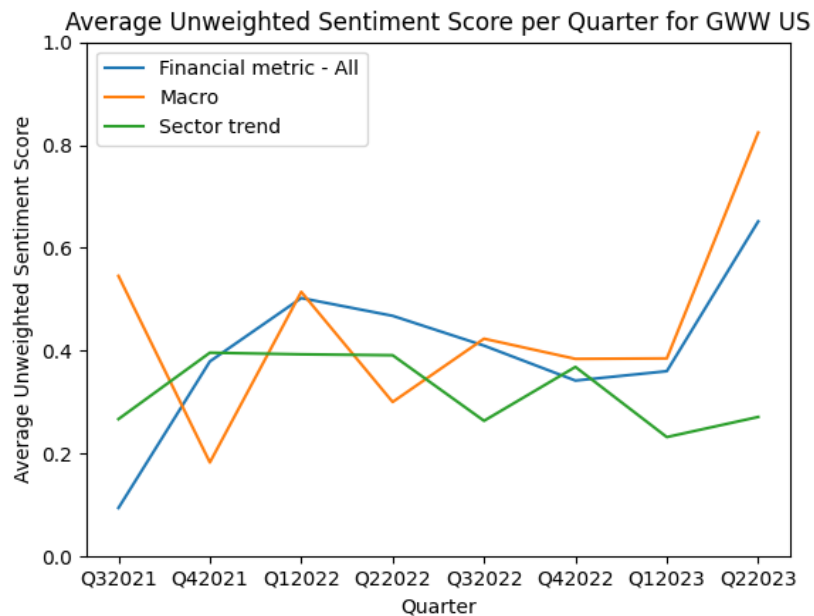
BAC US Sector Occurrences



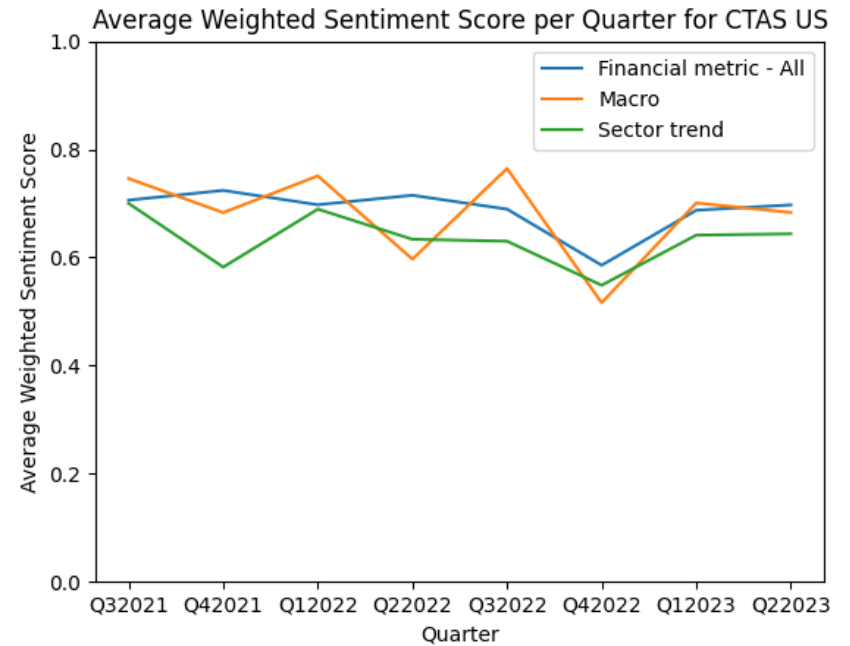
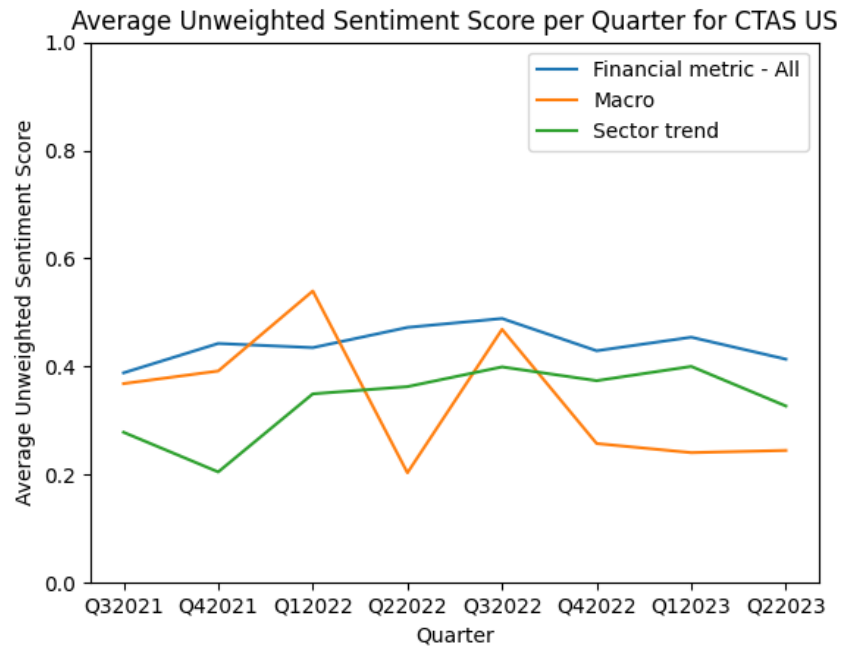
Raw score



# Graphs- WW Grainger Inc. (GWW US)

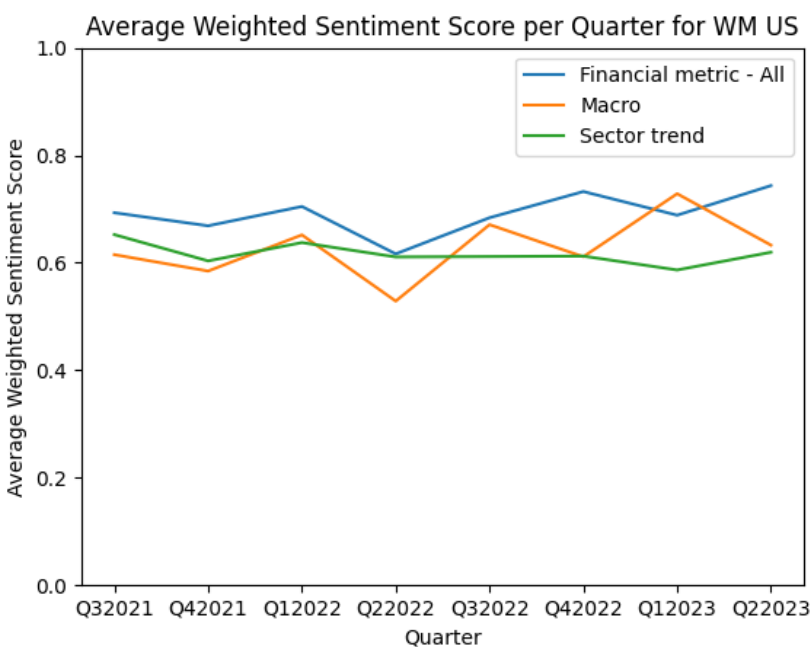
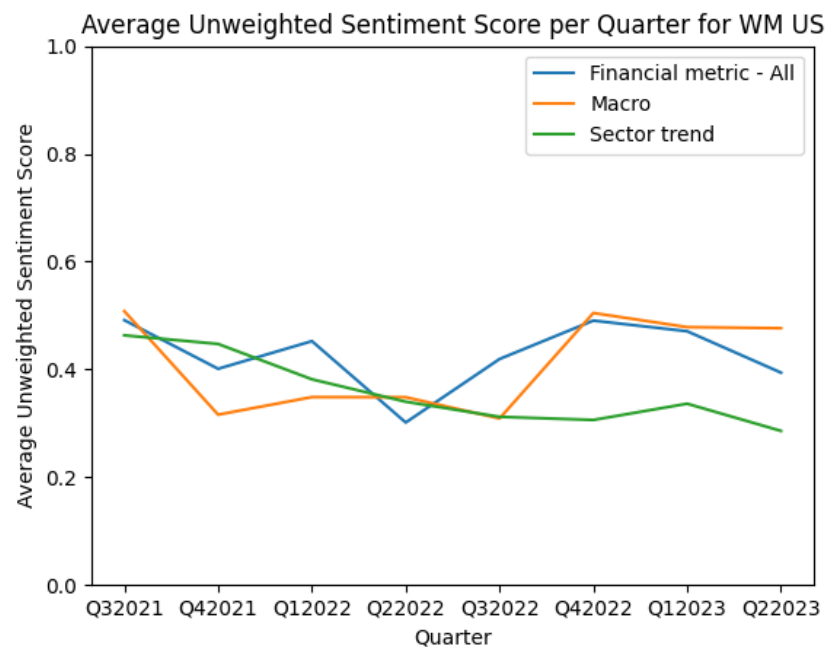


# Graphs- Cintas Corp (CTAS US)

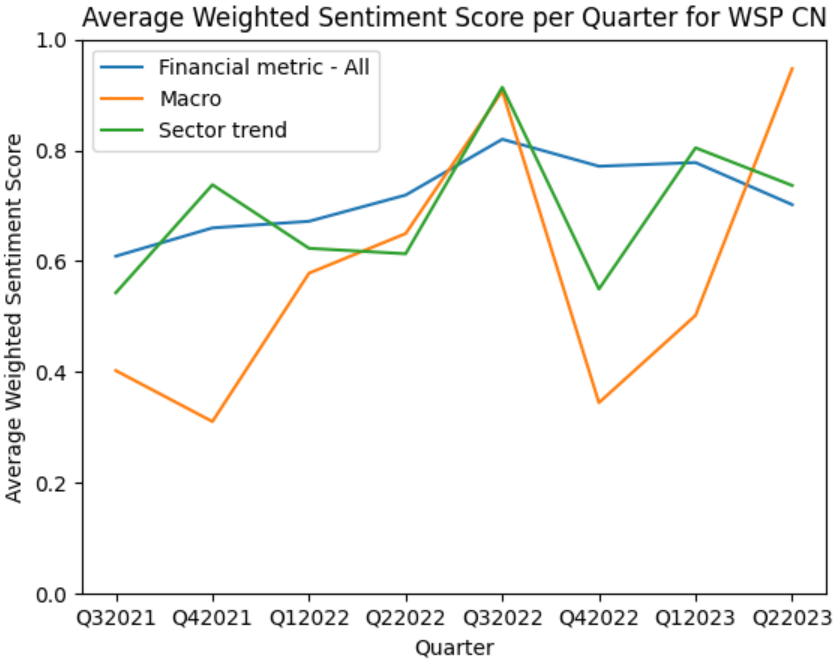
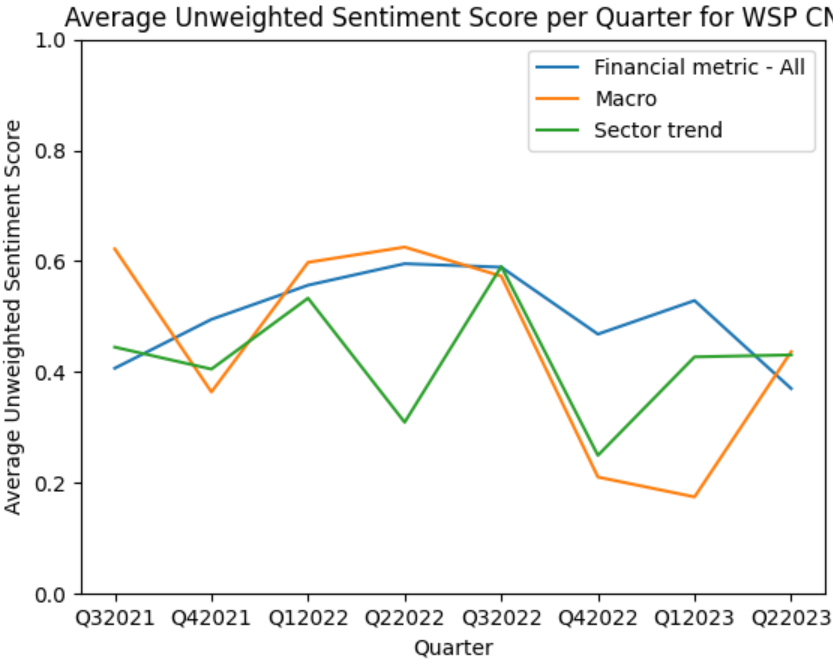




# Graphs- Waste Management (WM US)



# Graphs- From WSP Global Inc. (WSP CN)



# UConn Sentiment Data Manager

## Info Drill Down

Yahoo Ticker: Period: Category: 

Paragraph	Weighted Sentiment	Score	Keyword	Category
So those are a couple of things. Again, we're consistent with guidance. We are – we believe we're on track for the guidance we gave for the full-year of mid-single digits and we are looking forward to generating the efficiencies that come out of this spend and put a dent in our structural expense base over time.	0.828	0.923	Guidance	Financial metric - All

# Model

- BERT (Bidirectional Encoder Representations from Transformers)
  - Developed by Google AI, Industry standard
  - Used by major firms like Bloomberg for text analysis.
  - Pretrained on large-scale unlabeled data: Book Corpus & English Wikipedia
  - Reads context in both directions for deeper language comprehension
  - Finetune for specific tasks --> Sentiment Analysis

**UConn Sentiment Data Manager**

**Model Control**

Note: Each task is for a single ticker. Currently, all conference calls are processed from the ticker.

**Step 1. Input**

Upload Conference Calls

Ticker: BAC US

**Step 2. Settings**

Keyword List: None

Include Generic Keyword List: ☒

**Step 3. Confirm and Output**

Kind output:

Submit for Processing

Create Task Failed:

# Training

- Custom SSGA Training
  - Model can be trained on specific term
  - Industry Specific customizable weights
- Example: the word "reserves" can be trained as negative in the banking sector, despite being neutral in general sentiment analysis

sentence string · lengths	label class label
	3 classes
According to Gran , the company has no plans to move all production to Russia , although that is where the company is growing .	1 neutral
Technopolis plans to develop in stages an area of no less than 100,000 square meters in order to host companies working in computer technologies and telecommunications , the statement sai...	1 neutral
The international electronic industry company Elcoteq has laid off tens of employees from its Tallinn facility ; contrary to earlier layoffs the company contracted the ranks of its office...	0 negative
With the new production plant the company would increase its capacity to meet the expected increase in demand and would improve the use of raw materials and therefore increase the...	2 positive

# Roadmap

- **Data Acquisition**
  - Collect audio & text from financial conference calls
- **Data Preparation**
  - Clean data
- **Model Training**
  - Train using cleaned & segmented data
- **Model Evaluation**
  - Assess the model's performance using validation sets in addition to manual testing.
- **Adversarial Training**
  - Strengthen model robustness by training with Adversarial examples
- **Continuous Learning**
  - Regularly update the model with fresh data to adapt to evolving