Public Sentiment Toward the NFL

Can it predict weekly Fantasy Football outcomes?

- •Ryan Timbrook
- Diego Vales
- David Madsen



Data Science to help you win...



Using Data Science to Help Win Fantasy Football Games



COMMENTS 8



Social media and the rise of fantasy sports



How Top Fantasy Draft Companies Use Social Media

Fantasy draft sites emerged with the rise in fantasy sports popularity particularly, the popularity of fantasy on social media.

Fantasy Insights with Watson

Use AI to make better decisions in Fantasy Football

How it works

ESPN Fantasy Insights draws upon the latest in machine learning techniques to turn unstructured data into valuable insights. Nearly 10 million players rely on the combined resources of Watson Discovery and Watson OpenScale running on the IBM Cloud to give them a competitive edge.







<u>Get the latest updates on</u>

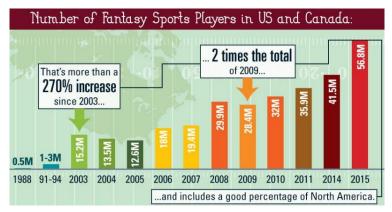




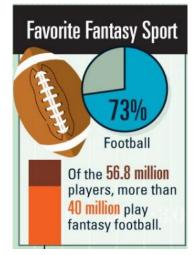
Fantasy Football Industry

Fantasy Football is an \$18.6 BILLION Market.









Average Time Spent Consuming Sports



Average Time Spent Consuming Fantasy Sports 9 hours per week

... Mapping Public Opinion To Weekly NFL Schedule ...



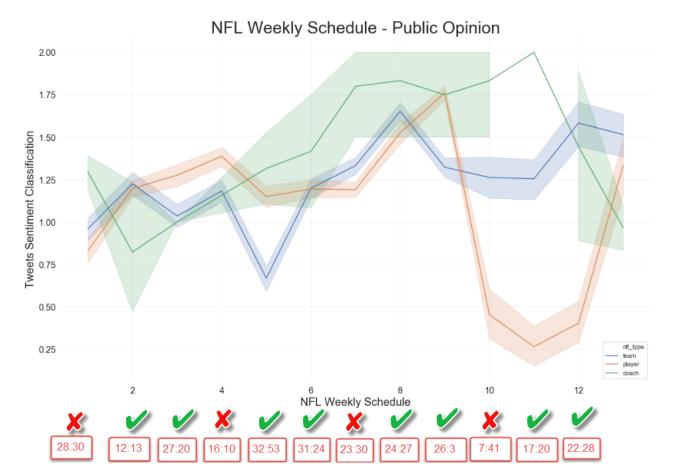












Team: Houston Texans

Coach: Bill O'Brian

Player: Deshaun Watson



VEDAR

- 1. positive sentiment: compound score >= 0.05
- 2. neutral sentiment: (compound score > -0.05) and (compound score < 0.05)
- 3. negative sentiment: compound score <= -0.05

compound score metric

1: Obtaining Twitter Data

Three Approaches Taken:

- Steaming API
 - Limited to 7 days data
- Full Archive Historical Search
 - Weekly search starting from NFL schedule week 1 to week 13
 - 10,000+ Tweets (Team, Coach, Player)
- Public Twitter Data Sets Pre-labeled
 - -Twitter US Airline Sentiment Analyze how travelers in February 2015
 - -First GOP Debate Twitter Sentiment Analyze tweets on the first 2016 GOP Presidential Debate

Usage numbers are updated at regular intervals but are not updated instantaneously. Graphs and data points should be accurate and updated to reflect actual product usage within one minute. All times are in UTC.

Current Month ~





Lesions Learnt

Request Usage Burn...

API Pagination Lesions
Learned...

Need for Search Operational Parameter Tuning

Use case for saving raw data to file — ran out of requests before report was written

Twitter API Wrappers...
--- Twython – doesn't have documentation on how to use premium vs. standard free

3: Twitter Data Sentiment Classification

4000

3000

8 ₂₀₀₀

mayfield lamar

VEDAR Classification on NFL Twitter Data

Polarity Scoring

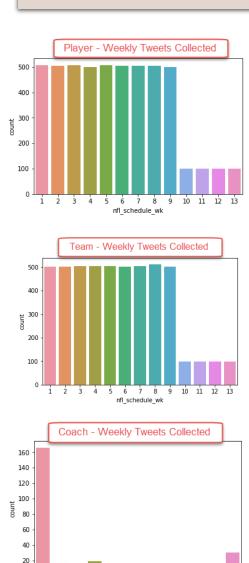
1.2

1.1

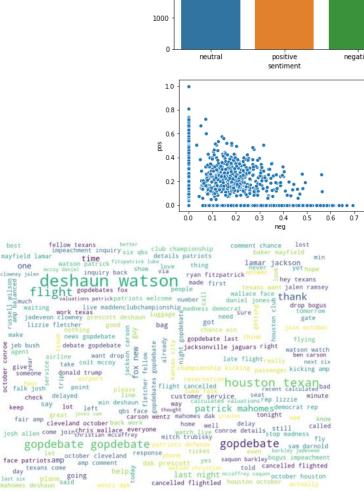
1.0

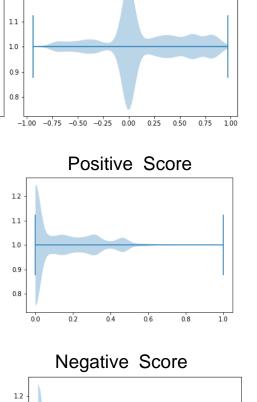
0.9

0.8



nfl_schedule_wk





Compound Score

4: Sentiment Classification Model Results

50,000 Labeled Tweets

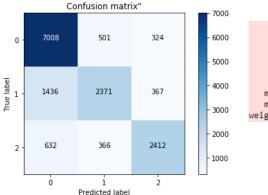
Best Fit Model: 77% Accuracy

Classification Type: LinearSVM

Vectorizer: TfidfVectorizer

N-Gram: Bigram

Binary: True



	precision	recall	f1-score	support
negative	0.77	0.89	0.83	7833
neutral	0.73	0.57	0.64	4174
positive	0.78	0.71	0.74	3410
micro avg	0.76	0.76	0.76	15417
macro avg	0.76	0.72	0.74	15417
weighted avg	0.76	0.76	0.76	15417

Classification Trials



Model_Type	Vectorizer	▼ N_Gram	Cross_Fold	Prediction_Accuracy	Total_Prediction_Points	Test_Recall_Score_Avg	Test_Precision_Score_Av	Train_Recall_Score_Av	Train_Precision_Score_Av	Total_Build_Time	Total_Predict_Time	Confusion_Matri:
												[[7008 501 324]
												[1436 2371 367]
inearSVM	tfidf	bigram	10	76.48	15417	0.7223	0.7599	0.87	0.9029	21.5007	0.0028	[632 366 2412]]
												[[1611 206 116]
												[260 717 93]
inearSVM	count	unigram	10	76.45	3855	0.7393	0.7481	0.8745	0.8889	34.694	0.0018	[128 105 619]]
												[[1585 258 100]
												[255 701 81]
inearSVM	count	bigram	10	75.77	3855	0.7383	0.7426	0.941	0.9457	62.9269	0.0011	[132 108 635]]
												[[6905 510 343]
												[1496 2349 390]
inearSVM	tfidf	unigram	10	75.57	15417	0.7152	0.7556	0.8259	0.8682	11.2072	0.002	[692 335 2397]]
												[[6305 1045 461]
			40	7450	45445	0.7040	0.7050	0.0540	0.0555	50 5076	0.0044	[1034 2671 424]
inearSVM	count	bigram	10	74.53	15417	0.7243	0.7268	0.9643	0.9656	50.5876	0.0044	[535 428 2514]]
												[[6316 977 470] [1123 2698 384]
inearSVM		unigram	10	74.44	15417	0.7142	0.7177	0.9141	0.922	38.1779	0.0023	[544 442 2463]]
inearsvivi	count	unigram	10	74.44	15417	0.7142	0./1//	0.9141	0.922	38.1779	0.0023	[[6304 985 532]
												[1050 2660 410]
inearSVM	count	unigram	10	74.31	15417	0.7168	0.7191	0.9143	0.9212	31.2185	0.0045	[566 418 2492]]
IIIeai 3 v Ivi	Count	unigram	10	74.31	15417	0.7100	0.7191	0.9143	0.9212	31.2103	0.0043	[[7299 283 193]
												[1889 1936 305]
Multinomial Naive Bave	es count	bigram	10	72.89	15417	0.6698	0.7502	0.746	0.8294	0.4043	0.006	[1154 355 2003]]
violenionioi_Noive_buy	Count	Digitatii	10	72.05	13117	0.0050	0.7302	0.7 40	0.0254	0.1013	0.000	[[7102 343 304]
												[1893 1893 443]
Aultinomial Naive Baye	es count	unigram	10	71.51	15417	0.6531	0.7245	0.7147	0.7924	0.2743	0.0036	[1070 340 2029]]
	200111	20.0		. 2102		2.0001	2.72.10	2.7 4 11	2.5521	JAET 10	2.3000	[[7133 320 292]
												[1923 1870 415]
Multinomial Naive Baye	es count	unigram	10	71.51	15417	0.6468	0.7238	0.7078	0.7903	0.2601	0.003	[1139 304 2021]]
				. 2102	27.00	2.0100	2.7200	2.7070				(

Predicting Player Outcomes from Tweets



DATA SCIENCE AT THE ISCHOOL AT SYRACUSE UNIVERSITY

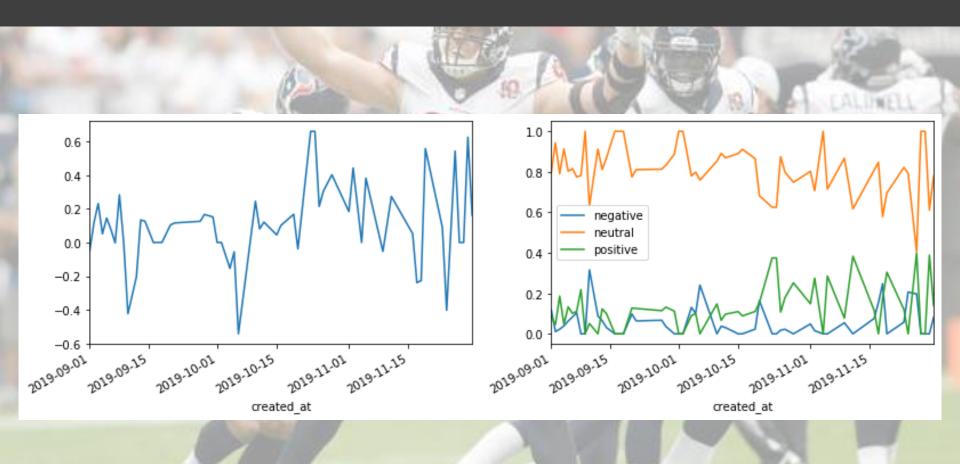
ARCHITECTURE

ACQUISITION

ANALYTIC

ARCHIVE

Tweet Sentiment Over Time



DATA SCIENCE AT THE ISCHOOL AT SYRACUSE UNIVERSITY

ARCHITECTURE

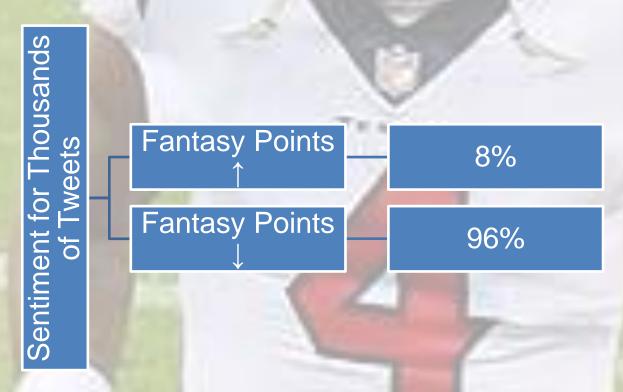
ACQUISITION

ANALYTICS

ARCHIVE

Predicting Player Outcomes from Tweet <u>Sentiment</u>

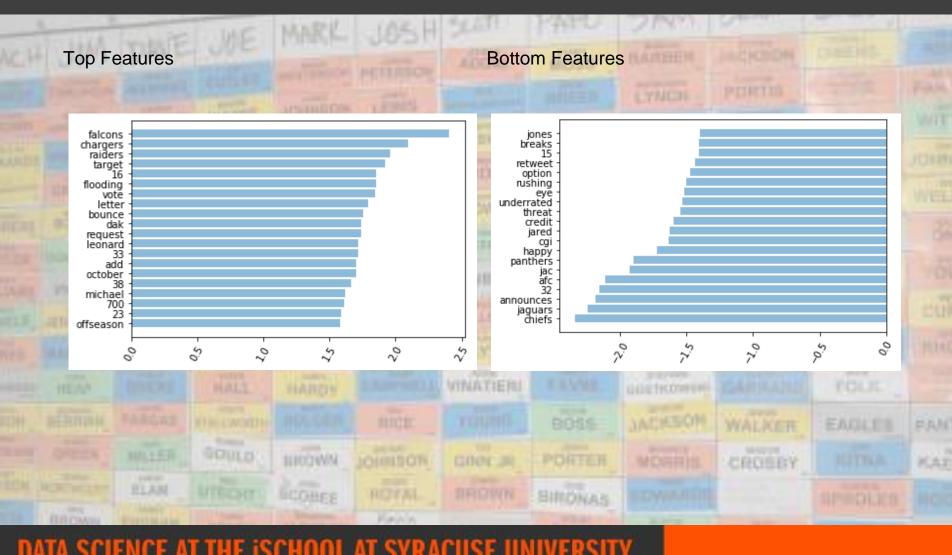
Case: Deshaun Watson Revisited



DATA SCIENCE AT THE ISCHOOL AT SYRACUSE UNIVERSITY

ARCHITECTURE ACQUISITION ANALYTICS ARCHIVE

Feature Importance



ARCHITECTURE **ACQUISITION**

ANALYTICS

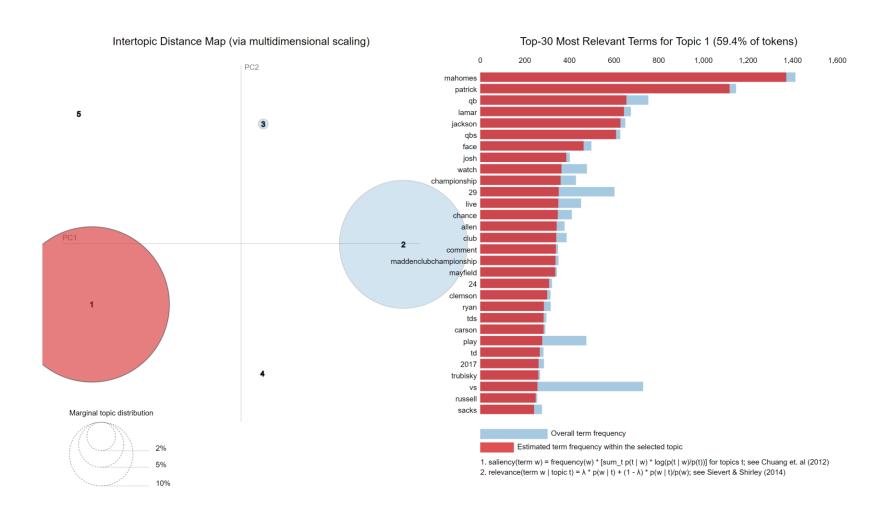
ARCHIVE

Topic Modeling

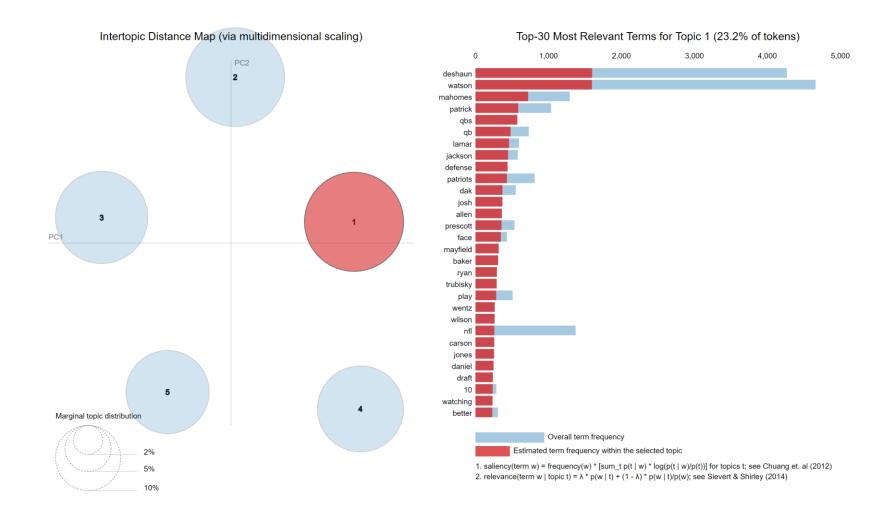
- Tweets
 - Player, Team, Coaches, Approaches
 - As files
 - As content
- Significance of emojis
 - Results
 - Included
 - Filtered



Topic Modeling - Files



Topic Modeling - Content





Question and Problem to Solve

Business Driver

- NFL revenue grew an estimated \$900 million to \$14 billion in 2017, in 2018 it generated about \$15 billion.
- Fantasy football and the spread of legalized sports betting across the U.S. promises to lock in fans and keep them focused on the game.
- As a Fantasy football player, how can Data Science help me make the most intelligence selections when deciding my weekly roster? "How do I win more?"

Problem to Solve

 Identify if real-time public popular opinion on NFL teams, players, and coaches is a predictor of their weekly fantasy football stats.

About the Data:

- Text data will be mined from three primary sources
- 1) Fantasy football websites like Yahoo Sports: https://sports.yahoo.com/fantasy/
- 2) Twitter Social Media API streams
- 3) Facebook Social Media API streams

Media Headlines



FAN FAVORITE

You might not have heard of Fanatics yet but it's taking over sports apparel one league at a time

Quartz

FANTASY FIGURES

How the \$7 billion US fantasy football industry makes its money in 2017

The bizarre, multibillion-dollar industry of American fantasy sports

December 13, 2014 · Quartz



CHOOSE WISELY

Fantasy football now matters more than official NFL teams

Quartz



There's finally a great use for IBM's Watson: making fantasy football picks

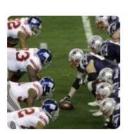
Quartz



REAL MONE

The two charts that explain why Yahoo is pushing deeper into fantasy sports

Quart



Why investors are pouring millions into fantasy sports

Quartz

Fantasy Football Industry

Q: How did fantasy go from weekender hobby to big-time business?

A: 2 words - DAILY FANTASY

In 2006, Congress passed the Unlawful Internet Gambling Enforcement Act (UIEGA), which banned online poker – deemed a game of chance – but left room for fantasy sports wagers – deemed a game of skill.



Notable Fantasy Football Brands

























DraftKings Promo Code and Review – Get 3 Free Months of RotoGrinders Premium for DraftKings