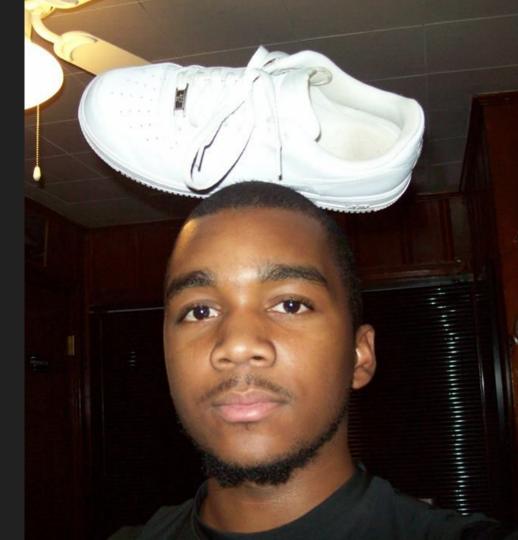
NIKE

What to do with the sneaker resell market?



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BACKGROUND KNOWLEDGE

- \$1 Billion annual market in sneaker resale
- 4% of sneakers bought on drop date are immediately resold
- Nike spends upwards of \$2.4 billion on marketing annually including a large social media budget
- Nike dominates the sneakerhead subculture with the intention of creating and maintaining brand loyalty

OBJECTIVES

- Is there a correlation between buzz created on Twitter and resale price on eBay?
- If so, can Nike use this information to predict when a shoe will be a successful drop?

DATA COLLECTION

Collect Twitter data to resemble public sentiment

- Tweepy/Python
- Pre-buzz: 2 days before the drop
- Post-buzz: 1 day after the drop
- Keyword search

2 Collect resale values from eBay

- Averaged twenty most recent sold listings
- Size of shoe ranged

3 Determine a statistic regression model

- Multiple linear regression
- DV: Resale Price
- IV: Pre buzz, Post buzz, Celebrity Endorsement, Original Price
- Statistically significant correlation found

DATA ANALYSIS

- Celebrity endorsement was NOT statistically significant
- To make model better
 - Removed celebrity endorsement
 - Took log of pre buzzand post buzz

SUMMARY OUTP	UT				
Regression	Statistics				
Multiple R	0.959686437				
R Square	0.920998058				
Adjusted R Squar	0.861746601				
Standard Error	18.18506932				
Observations	8				
ANOVA					
	df	SS	MS	F	Significance F
Regression	3	15420.9404	5140.313468	15.54388886	0.011389563
Residual	4	1322.786984	330.6967461		
Total	7	16743.72739			
	Coefficients	Standard Error	t Stat	P-value	Lower 95%
Intercept	-47.30283951	68.65487267	-0.688994643	0.528709657	-237.9193247
Oringial Price	1.785235883	0.353501217	5.050154847	0.0072303	0.80375916
Log Pre Buzz	61.6865989	13.7451903	4.487867942	0.010923665	23.52383258
Log Post Buzz	-69.14684892	20.12529406	-3.435818067	0.026394583	-125.0236231

MAIN CHALLENGES

1 | Finding data

- Deadstock per release
- Quantity sold on eBay
- Percentage of purchase shoes that are resold

2 | Rate Limitation

- Number of requests could not exceed 180
- Combat rate limitation by grabbing tweets by region with geocode
- 429 error, 15 minute wait

3 | Date Limitation

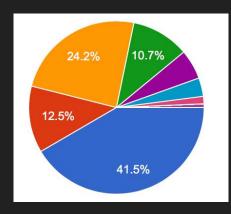
- Lack of time specificity with since and until parameters
- Pulling Tweet ID with desired created_at
- Historical data cannot be older than one week

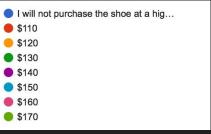
INSIGHTS

- Resale Market
 - 98% of shoe resale market is Nike brand
 - Nike maintains and generates brand loyalty through exclusivity
- Re-releases
 - Limited deadstock during initial release
 - Re-released shoes have different colors, patterns, etc
 - Same price during re-release

RECOMMENDATIONS

- Do NOT increase price before original drop
- Use model to predict popularity of shoe
 - \circ High buzz \rightarrow popularity
 - \circ Popularity \rightarrow high resale price
 - High resale price \rightarrow market for higher price
- Increase subsequent drop price by 10% of original price





SWOT ANALYSIS

Strengths

Simple model with data that encompasses many aspects of a shoe

Weaknesses

- Sample size of shoe drops
- Resell data from eBay

• Opportunities

 Collecting more data will lead to a more accurate model with the ability to include more parameters

• Threats

- Nike has a highly detailed model already in existence
- o Countless resources available online for sneakerheads

Questions?

