

Do IPOs offer a degree of market dominance?

- 1058 Initial Public Offerings in 2021
- IPOs close 2021 with over \$300 billion in market share for the year
- \$153.5 Billion in traditional offerings
- \$162 Billion in blank check companies (SPAC)
- Overall market performance at all time highs

What's the scoop?

- Fundamental data points can correlate to predict potential outcomes
 - Annual Financials, Offer Date, Industry, S&P Returns
- Using machine learning and neutral network models we can analyze these fundamentals
 - Voting classifier, PyTorch, Logistic Regression, SVC, Adaboost, Xgboost, Random Forest
- Can a lucrative trading strategy be developed?
 - Getting in on a stock pre-IPO usually comes with the caveat that it cannot be traded for a period of time.

DATA PREP

IPOScoop.com

306 Stocks

Yahoo Finance *Plus*

Annual Income Statement

Balance Sheet

Cash Flows

	Symbol	Industry	Offer Date	Shares (millions)	Offer Price
1	ABOS	Health Care	7/1/2021	10	\$16.00
2	ABSI	Health Care	7/22/2021	12.5	\$16.00
3	ACHL	Health Care	3/31/2021	9.8	\$18.00
4	ACT	Financials	9/16/2021	13.3	\$19.00
5	ACVA	Other	3/24/2021	16.6	\$25.00
6	ACXP	Health Care	6/25/2021	2.5	\$6.00
7	AIP	Technology	10/27/2021	5	\$14.00
8	AIRS	Health Care	10/29/2021	7	\$11.00
9	AKA	Consumer Goods	9/22/2021	10	\$11.00
10	ALHC	Health Care	3/26/2021	27.2	\$18.00

	name	12/31/2020	12/31/2019
1	TotalAssets	44,429,000	7,322,000
2	CurrentAssets	44,429,000	7,178,000
3	CashCashEquivalen	43,777,000	6,552,000
4	CashAndCashEquiv	43,777,000	6,552,000
5	Receivables	109,000	30,000
6	OtherReceivables	109,000	30,000
7	PrepaidAssets	91,000	49,000
8	OtherCurrentAssets	452,000	547,000
9	TotalNonCurrentAssets	0	144,000
10	NonCurrentDeferred	0	0
11	OtherNonCurrentAs		144,000
12	TotalLiabilitiesNetMin	63,020,000	18,742,000
13	CurrentLiabilities	6,367,000	1,342,000
14	PayablesAndAccrue	864,000	728,000
15	Payables	531,000	223,000
16	AccountsPayable	531,000	223,000
17	CurrentAccruedExp	333,000	505,000
18	PensionandOtherPo	0	0
19	OtherCurrentLiabilities	5,503,000	614,000
20	TotalNonCurrentLiabi	56,653,000	17,400,000
21	PreferredSecurities	56,653,000	17,400,000
22	TotalEquityGrossMino	-18,591,000	-11,420,000
23	StockholdersEquity	-18,591,000	-11,420,000
24	CapitalStock	0	0
25	CommonStock	0	0
26	AdditionalPaidInCap	8,374,000	8,220,000
27	RetainedEarnings	-26,965,000	-19,640,000
28	TotalCapitalization	-18,591,000	-11,420,000
29	CommonStockEquity	-18,591,000	-11,420,000
30	NetTangibleAssets	-18,591,000	-11,420,000
31	WorkingCapital	38,062,000	5,836,000
32	InvestedCapital	-18,591,000	-11,420,000
33	TangibleBookValue	-18,591,000	-11,420,000
34	ShareIssued	38,651,795	36,985,129
35	OrdinarySharesNumber	38,651,795	36,985,129

	name	ttm	12/31/2020	12/31/2019
1	TotalRevenue	329,000	1,436,000	1,697,000
2	OperatingRevenue	329,000	1,436,000	1,697,000
3	OperatingExpense	12,839,000	9,348,000	9,502,000
4	SellingGeneralAndA	5,181,000	1,351,000	926,000
5	GeneralAndAdminist	5,181,000	1,351,000	926,000
6	OtherGandA	5,181,000	1,351,000	926,000
7	ResearchAndDevelo	7,658,000	7,997,000	8,576,000
8	OperatingIncome	-12,510,000	-7,912,000	-7,805,00
9	NetNonOperatingInter	22,000	1,000	45,000
10	InterestIncomeNonO	45,000	1,000	45,000
11	OtherIncomeExpense		586,000	-147,000
12	GainOnSaleOfSecurity		586,000	-147,000
13	Pretaxincome	-93,012,000	-7,325,000	-7,907,000
14	NetIncomeCommonS	-93,598,000	-7,911,000	-7,686,000
15	Netincome	-93,012,000	-7,325,000	-7,907,00
16	NetIncomeIncluding	-93,012,000	-7,325,000	-7,907,00
17	NetIncomeContinuo	-93,012,000	-7,325,000	-7,907,00
18	OtherunderPreferred		586,000	-221,00
19	DilutedNIAvailtoCom	-93,598,000	-7,911,000	-7,686,00
20	BasicEPS		-0.28	-0.26
21	DilutedEPS		-0.28	-0.26
22	BasicAverageShares		28,651,796	28,651,79
23	DilutedAverageShares		28,651,796	28,651,79
24	TotalOperatingIncome	-12,510,000	-7,912,000	-7,805,00
25	TotalExpenses	12,839,000	9,348,000	9,502,00
26	NetIncomeFromConti	-93,012,000	-7,325,000	-7,907,00
27	NormalizedIncome	-12,441,000	-7,911,000	-7,760,00
28	Interestincome	45,000	1,000	45,00
29	NetInterestincome	22,000	1,000	45,00
30	EBIT	-93,576,000	-7,912,000	-7,805,00
31	EBITDA	-93,576,000		
32	NetIncomeFromConti	-93,012,000	-7,325,000	-7,907,00
33	TotalUnusualitemsEx	-80,571,000	586,000	-147,00
34	TotalUnusualitems	-80,571,000	586,000	-147,00
35	NormalizedEBITDA	-13,005,000	-8,498,000	-7,658,00
36	TaxRateForCalcs	0	0	
37	TaxEffectOfUnusualit	0	0	

	name	ttm	12/31/2020	12/31/2019
1	OperatingCashFlow	-17,179,000	-7,450,000	-6,818,000
2	CashFlowFromConti	-17,179,000	-7,450,000	-6,818,000
3	NetIncomeFromCon	-93,012,000	-7,325,000	-7,907,00
4	Operating Gains Losses		-586,000	147,00
5	GainLossOnInvest		-586,000	147,00
6	StockBasedCompen	595,000	154,000	174,00
7	ChangeInWorkingCa	-5,328,000	307,000	768,00
8	ChangeInReceivables	108,000	-79,000	200,00
9	ChangeInPrepaidAs	-4,705,000	53,000	6,00
10	ChangeInPayables	-1,240,000	189,000	642,00
11	ChangeInPayable	-1,072,000	308,000	110,00
12	ChangeInAccount	-1,072,000	308,000	110,00
13	ChangeInAccruedE	-168,000	-119,000	532,00
14	ChangeInOtherCurr	509,000	144,000	-80,00
15	FinancingCashFlow	245,131,000	44,675,000	6,239,00
16	CashFlowFromConti	245,131,000	44,675,000	6,239,00
17	NetPreferredStockIs	74,706,000	44,675,000	6,232,00
18	PreferredStockIssu	74,706,000	44,675,000	6,232,00
19	ProceedsFromStock	1,866,000	0	7,00
20	EndCashPosition	135,802,000	43,777,000	6,552,00
21	ChangesInCash	133,843,000	37,225,000	-579,00
22	BeginningCashPosition	1,959,000	6,552,000	7,131,00
23	IncomeTaxPaidSuppl	0	0	
24	InterestPaidSupplem	0	0	
25	IssuanceOfCapitalStock	243,265,000	44,675,000	6,232,00
26	FreeCashFlow	-17,193,000	-7,450,000	-6,818,00

DATA

PREP

```
# Use Yahoo Finance data to measure returns for above-mentioned 100 day period
for ticker in ipo df.index:
                                                                                                                               bz_ipo_df['100D Y/N'].value_counts()
    data = yf.download('SPY',ipo_df.loc[ticker,'T-90D'],ipo_df.loc[ticker,'T-10D'],progress=False)
   spy_return = (data.iloc[-1,-2] - data.iloc[0,-2])/data.iloc[0,-2]
   ipo_df.loc[ticker,'SPY 90D Return'] = spy_return
                                                                                                                               0.0
                                                                                                                                          214
#Use Yahoo Finance data to measure returns for +100 days
                                                                                                                               1.0
                                                                                                                                           92
   data2 = yf.download(ticker,ipo_df.loc[ticker,'Offer Date'],ipo_df.loc[ticker,'T+100D'], progress=False)
                                                                                                                               Name: 100D Y/N, dtype: int64
    stock_return = (data2.iloc[-1,-2] - data2.iloc[0,-2])/data2.iloc[0,-2]
    ipo_df.loc[ticker,'100 day Return'] = stock_return
                                                            # Use RandomOverSampler to balance data
```

X_resampled, y_resampled = ros.fit_resample(X_train, y_train)

ros = RandomOverSampler(random_state=1)

431 Features

201

Loop through columns and drop columns that have 0 for more than 20% of the data therein for column in df.columns[1:]: zeros = (df[column]==0).sum() if zeros >= 0.80*len(df.index): df.drop(columns = column, inplace = True) else: pass return df

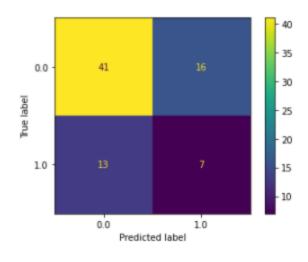
Feature Importance

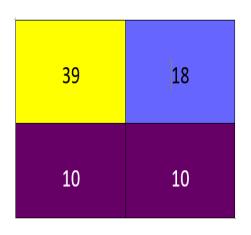
Offer Date	0.043421
InvestedCapital	0.013331
SPY 980 Return	0.013099
TotalLiabilitiesNetMinorityInterest	0.012707
CashAndCashEquivalents	0.011619
TotalRevenue	0.011036
ChangeInWorkingCapital	0.010986
CashFlowFromContinuingFinancingActivities	0.010420
OperatingRevenue	0.009763
RetainedEarnings	0.009601

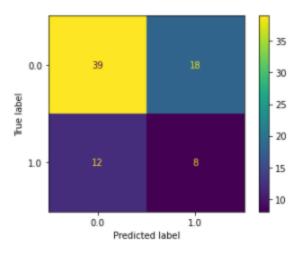
Xg Boost

Pytorch

Ada Boost







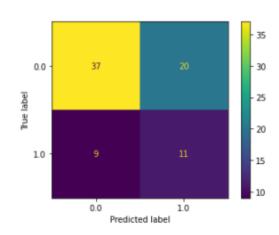
XGB Classification Report							
	precision	recall	f1-score	support			
0.0	0.76	0.72	0.74	57			
1.0	0.30	0.35	0.33	20			
accuracy			0.62	77			
macro avg	0.53	0.53	0.53	77			
weighted avg	0.64	0.62	0.63	77			

PyTorch Model Classification Report				
	precision	recall	f1-score	support
0.0	0.80	0.68	0.74	57
1.0	0.36	0.50	0.42	20
accuracy			0.64	77
macro avg	0.58	0.59	0.58	77
weighted avg	0.68	0.64	0.65	77

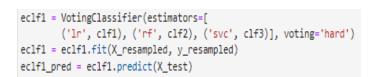
ADA + SVO	Cla	ssification	Report		
		precision	recall	f1-score	support
	0.0	0.76	0.68	0.72	57
	1.0	0.31	0.40	0.35	20
accur	racy			0.61	77
macro	avg	0.54	0.54	0.54	77
weighted	avg	0.65	0.61	0.62	77

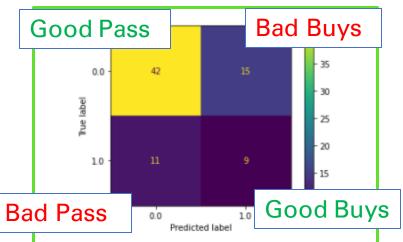
Voting Classifier with SVC, Random Forest, and Logistic Regression

```
clf1 = LogisticRegression(multi_class='multinomial', random_state=1)
clf2 = RandomForestClassifier(n_estimators=1000, random_state=1)
clf3 = SVC(probability = True, kernel = 'sigmoid', random_state=1)
```

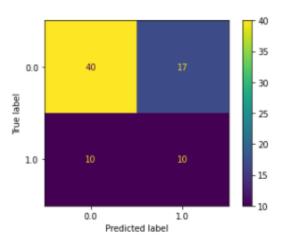


VotingClassif	ier Classifi	cation Re	port	
	precision	recall	f1-score	support
0.0	0.80	0.65	0.72	57
1.0	0.35	0.55	0.43	20
accuracy			0.62	77
macro avg	0.58	0.60	0.57	77
weighted avg	0.69	0.62	0.64	77

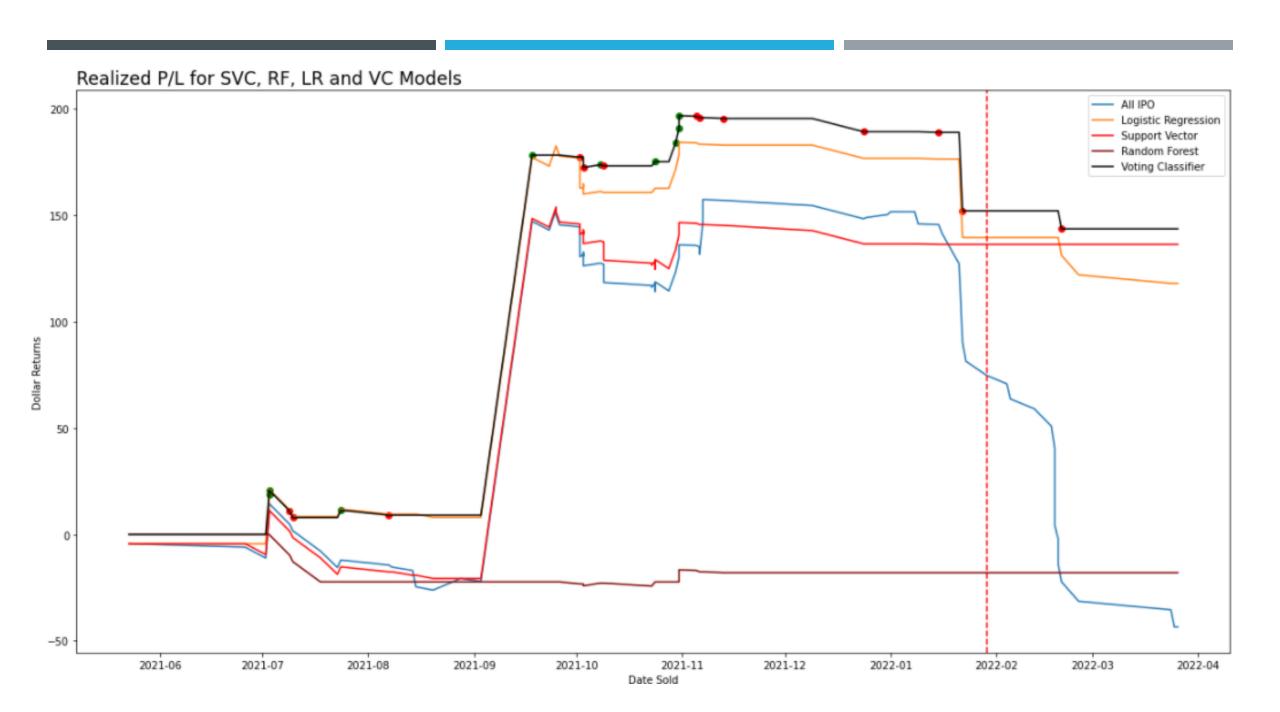




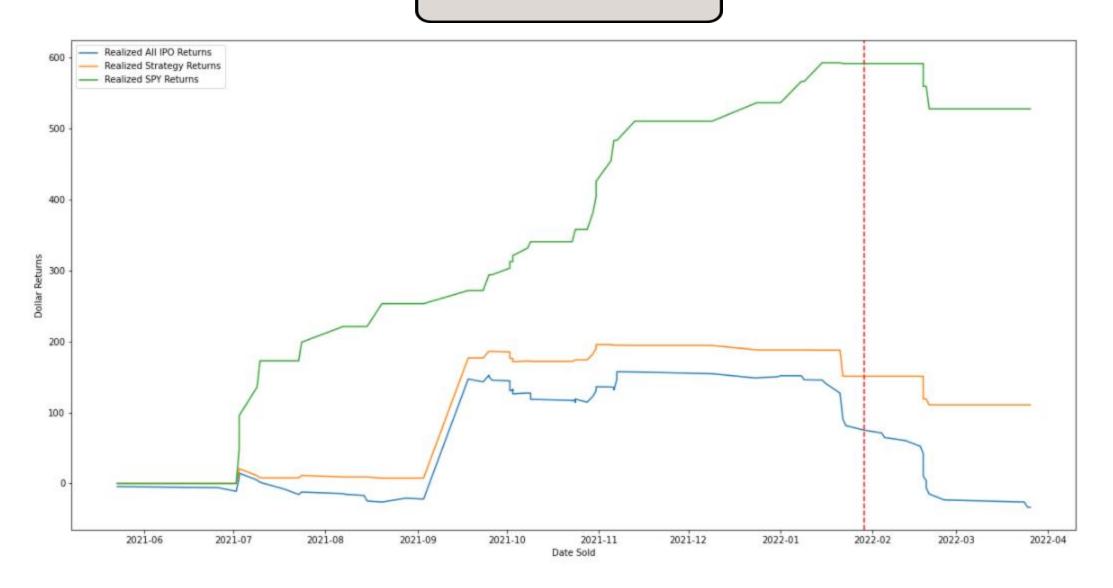
VotingCl	.assif	ier Classifio	ation Re	port	
		precision	recall	f1-score	support
	0.0	0.79	0.74	0.76	57
	1.0	0.38	0.45	0.41	20
accu	ıracy			0.66	77
macro	avg	0.58	0.59	0.59	77
weighted	avg	0.68	0.66	0.67	77
eclf3 =	('lr' votir flatt eclf	ingClassifi , clf1), (ng='soft', en_transfo 3.fit(X_re eclf3.pre	'rf', c weights rm=True sampled	lf2), ('s =[1,2,2],) , y_resam	vc', clf3)], pled)



VotingClassifier Classification Report precision recall f1-score support 0.0 0.80 0.70 0.75 57 1.0 0.37 0.43 20 0.50 77 accuracy 0.65 0.59 77 macro avg 0.59 0.60 0.69 77 weighted avg 0.65 0.66



Voting Classifier vs DCA SPY



QUESTIONS WE STILL HAVE

IDEAS FOR IMPROVEMENT

QUESTIONS YOU MAY HAVE

SUGGESTIONS?

