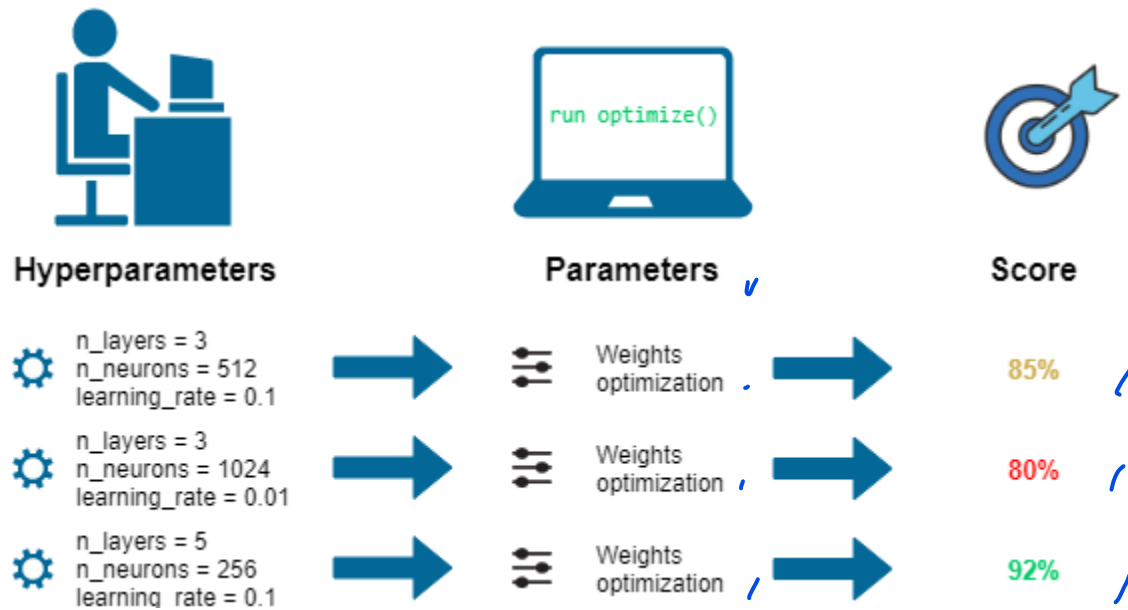




GridSearch Wrapup

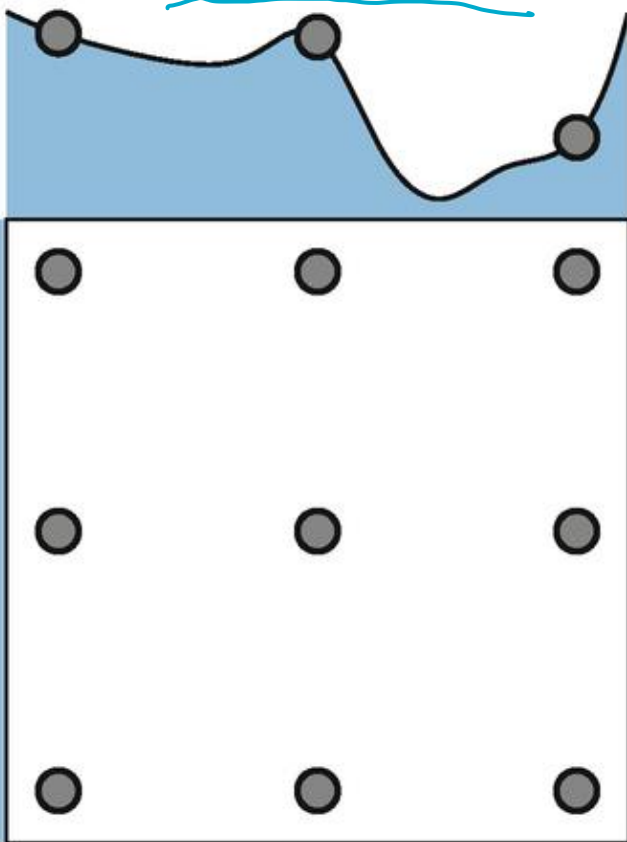




GridSearch Wrapup

Unimportant parameter

Grid Search

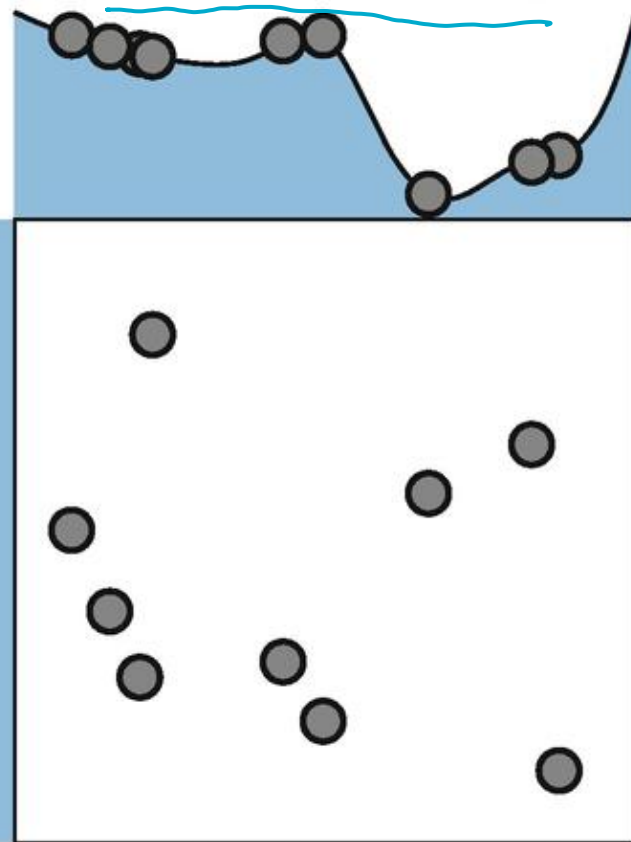


Important parameter

up step $\frac{1}{n}$ division
↓
 $\frac{1}{n}$ Prob
ex MacD 1, 2, 4,
5, 10, 1, 12

Random Search

Unimportant parameter



Important parameter





GridSearch Wrapup

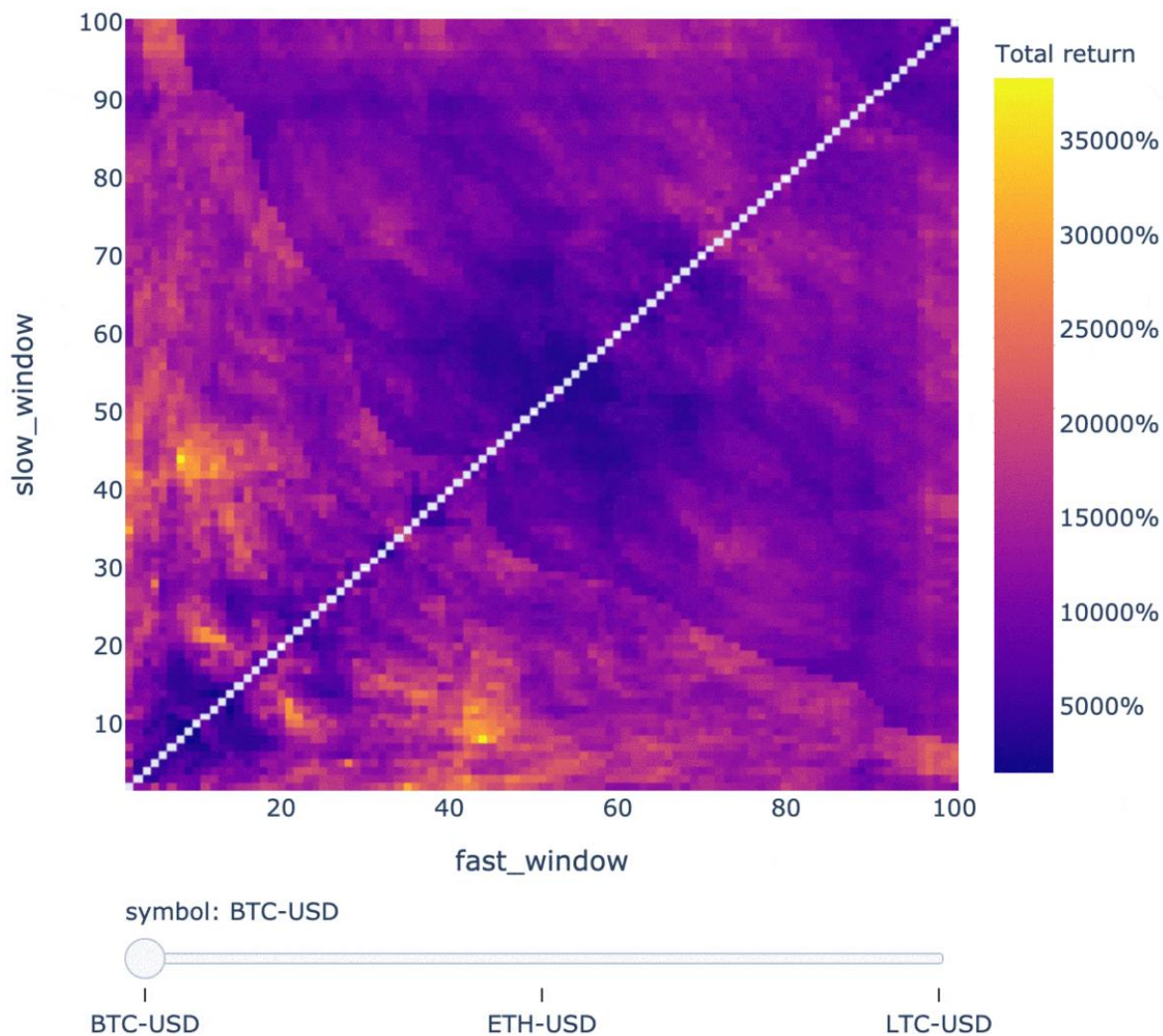
Macd Fast EMA	MACD Slow EMA	MACD SMA	Profit in pips
12	26	9	390
12	26	5	381
12	26	18	376
24	52	18	370
24	26	9	360
12	34	9	314
12	14	9	306
6	26	9	260
6	13	5	220
forex.in.rs			

20% modification ✓



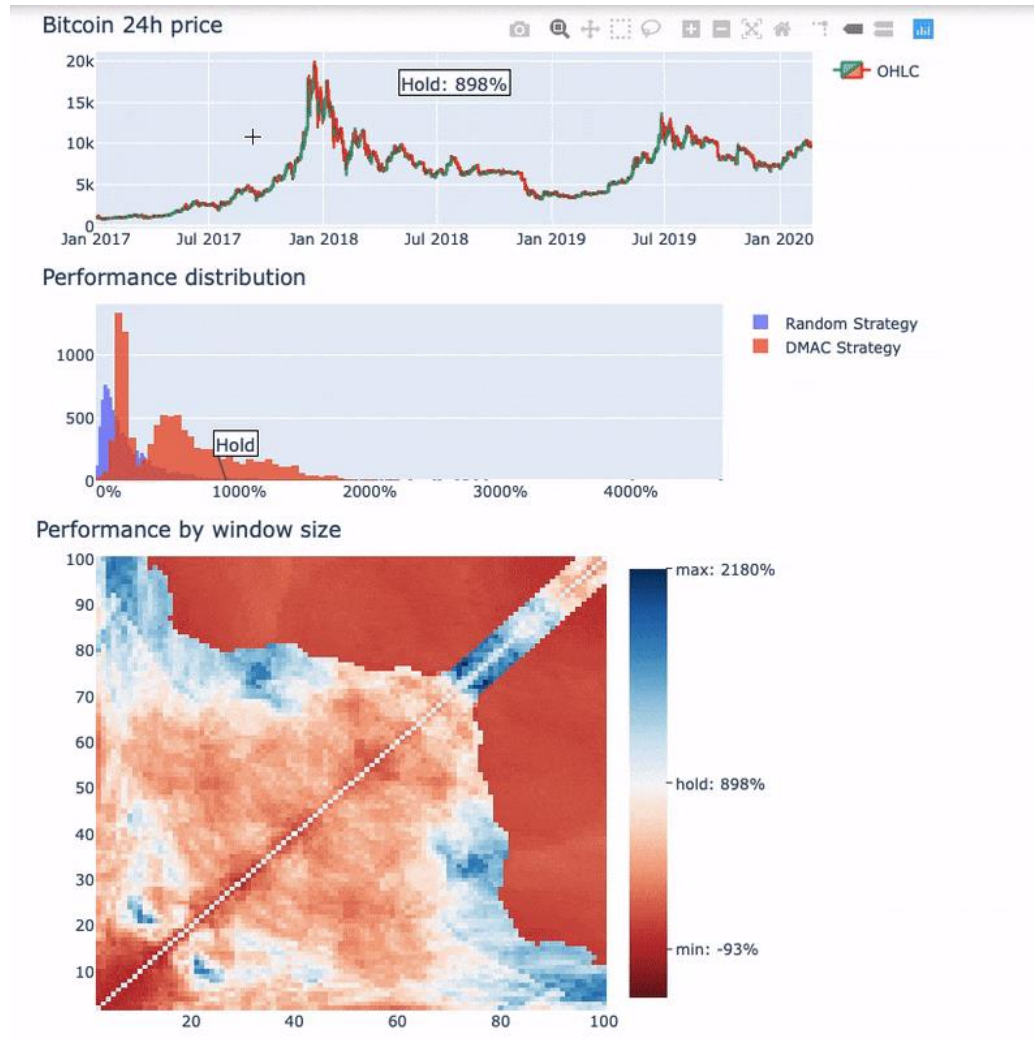


Backtesting vs Parameter Tuning





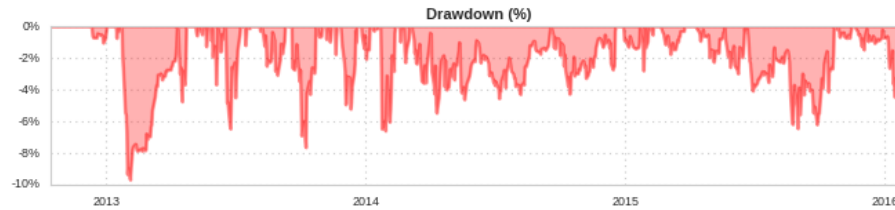
Backtest Wrapup





Backtest Wrapup

Sentiment Sentdex Strategy - Defence Stocks



Monthly Returns (%)												
2012	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	-0.0
2013	-7.0	2.4	4.7	7.8	9.1	1.5	11.9	1.2	6.9	5.7	6.4	4.2
2014	3.2	5.8	0.1	-1.4	0.3	-1.2	-0.2	2.3	1.1	-1.6	1.6	2.2
2015	0.8	4.7	1.5	-1.3	-0.6	-1.5	2.2	-3.3	-0.4	7.5	3.4	0.6
2016	-3.0	0.2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec



Curve vs. Benchmark		
Total Return	111%	34%
CAGR	25.45%	9.42%
Sharpe Ratio	1.69	0.75
Sortino Ratio	2.19	1.01
Annual Volatility	14.04%	13.21%
R-Squared	0.88	0.84
Max Daily Drawdown	9.69%	13.04%
Max Drawdown Duration	138	175
Trades per Year	24.6	

Trade	
Trade Winning %	57%
Average Trade %	3.75%
Average Win %	8.33%
Average Loss %	-2.27%
Best Trade %	52.78%
Worst Trade %	-7.46%
Worst Trade Date	TBD
Avg Days in Trade	0.0
Trades	81

Time	
Winning Months %	71%
Average Winning Month %	3.43%
Average Losing Month %	-1.80%
Best Month %	11.90%
Worst Month %	-7.04%
Winning Years %	60%
Best Year %	68.85%
Worst Year %	-2.76%

Συστάματα
on limit →
Drawdown



INVESTIC



Skill Trees

Bank 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100

Home Trader

Technical Analysis
+ Algo Trade

Advance Systematic
Trading Algorithm

Python

Quantitative
Analysis

Risk Parity
All Weather Portfolio

Black Litterman

Bank
/ Fund Mgmt

Portfolio Management
Mean Variance Optimization

Smart Beta
Factor Investing

Data Sci

Machine Learning

Trees

Deep Learning

Reinforcement
Learning

Deep Reinforcement
Learning

INVESTIC

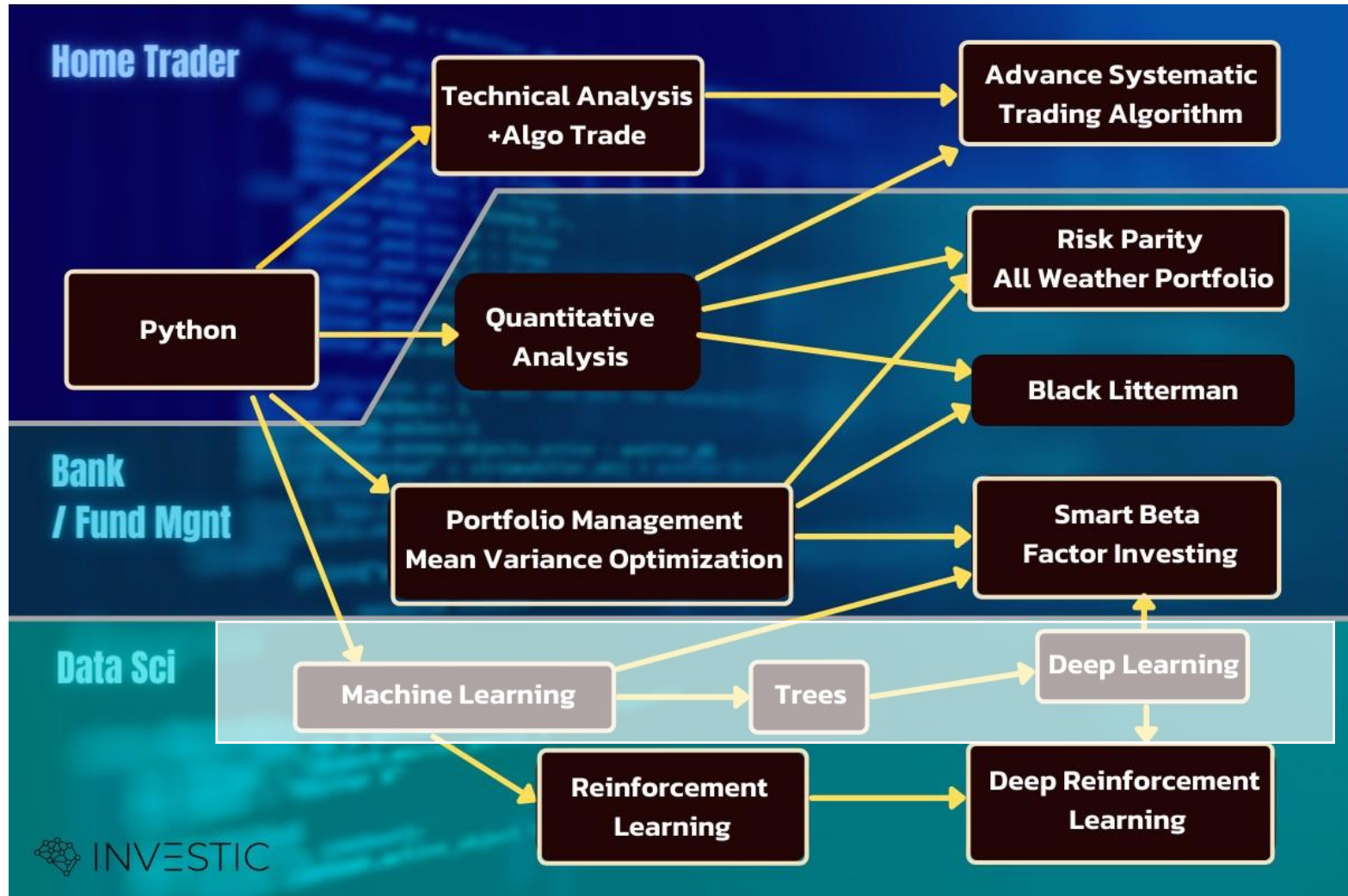
8. 20 04



INVESTIC

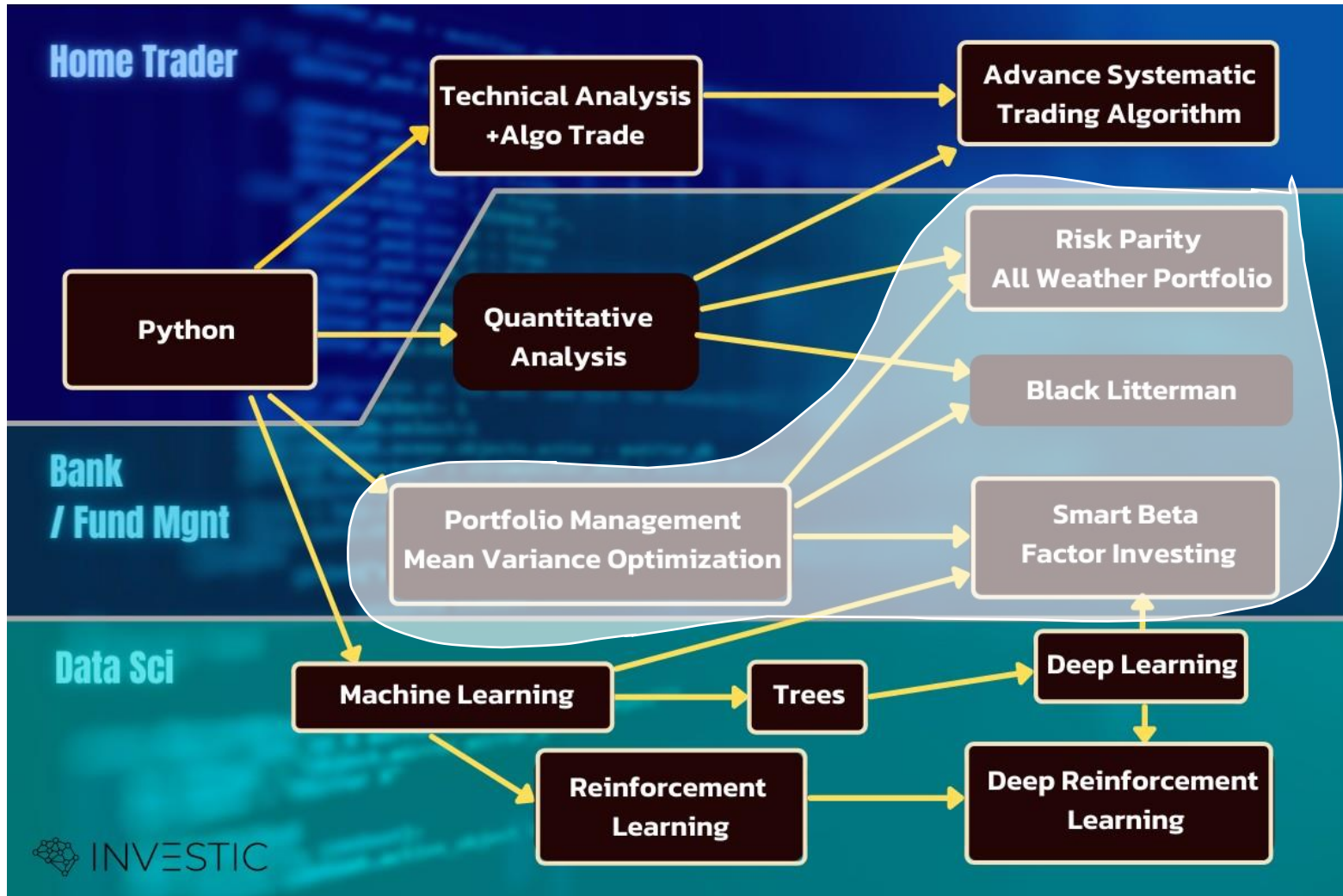


Machine Learning Trees



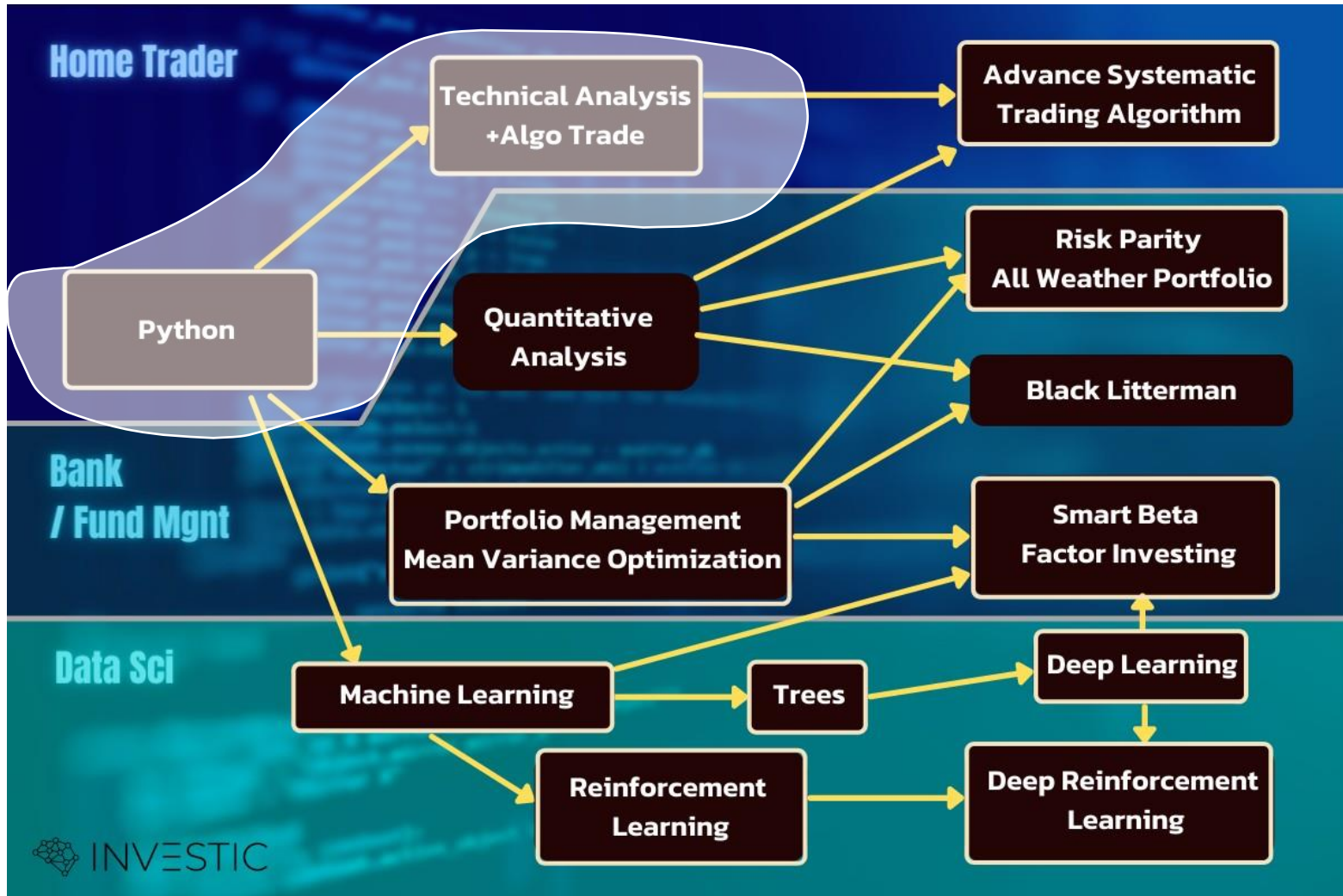


Portfolio Management Trees





Technical Analysis Trees

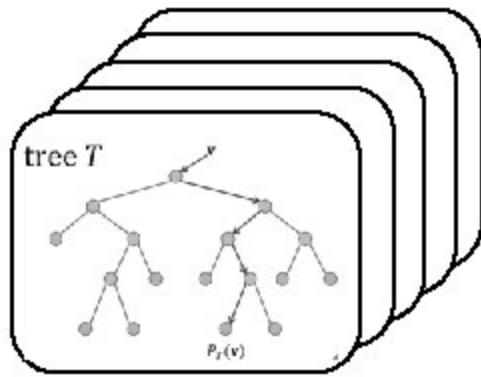




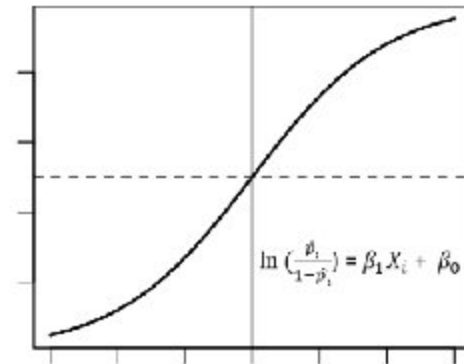


What do you think?

—

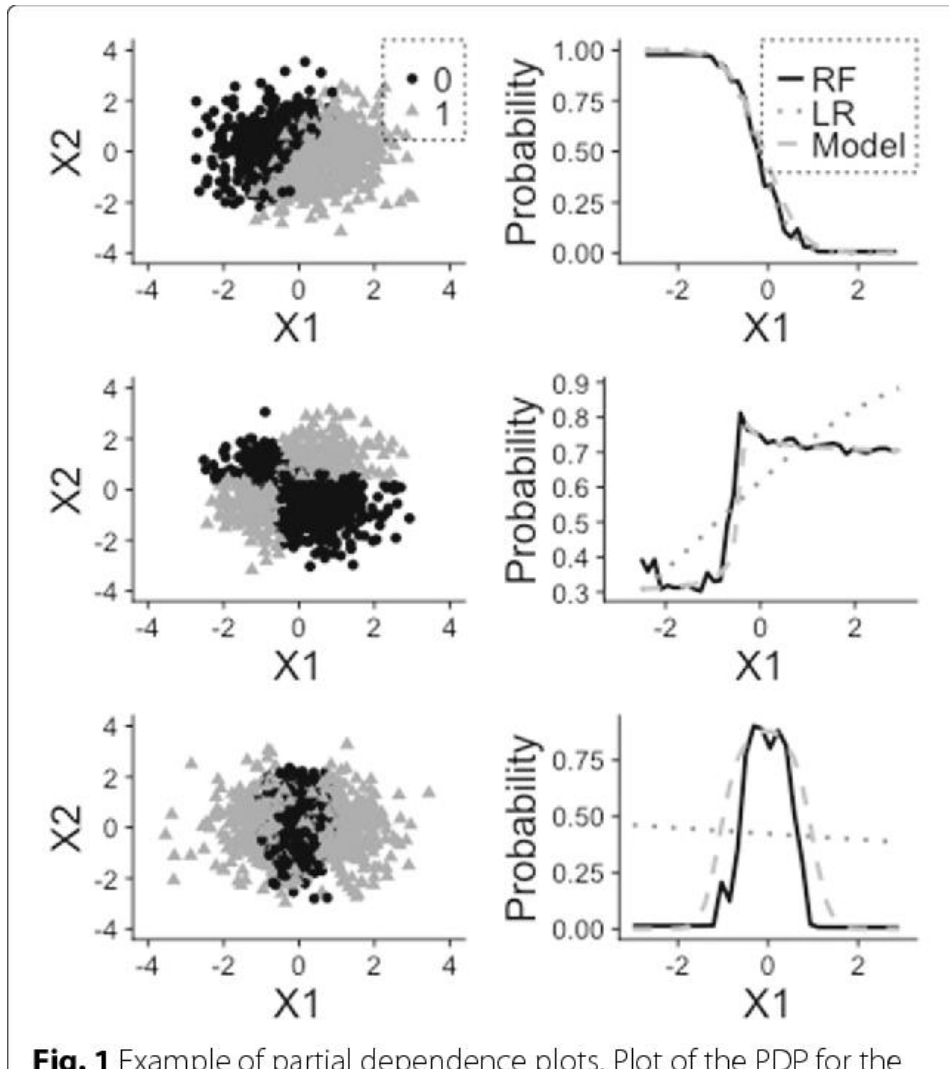


VS



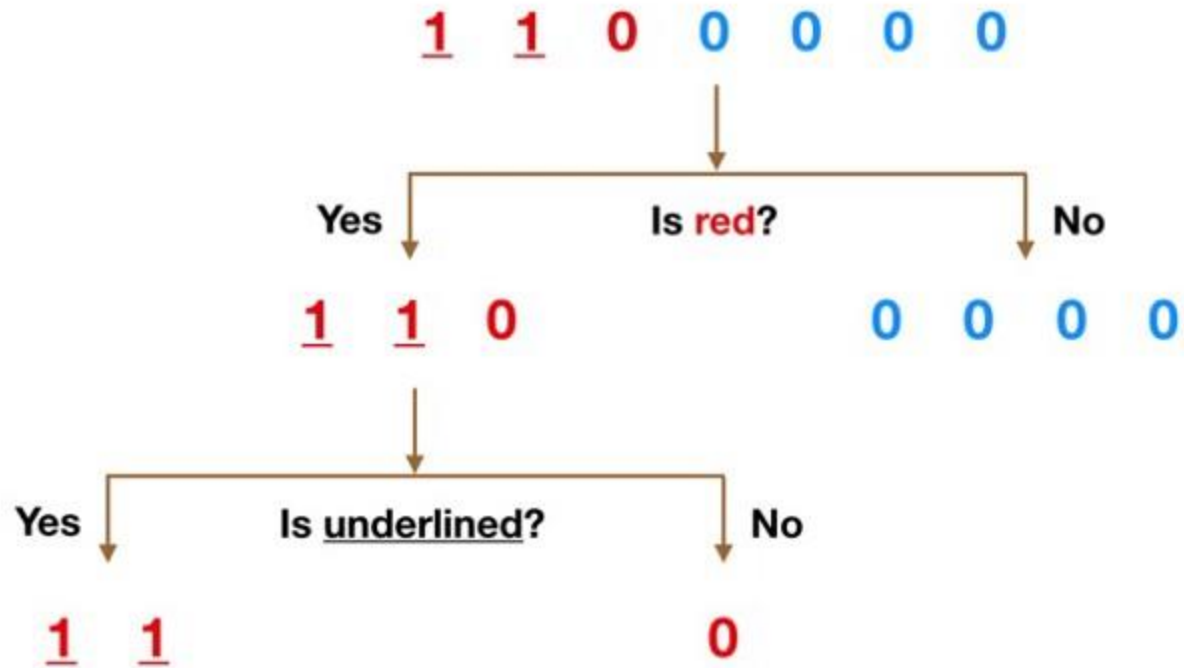


What do you think?





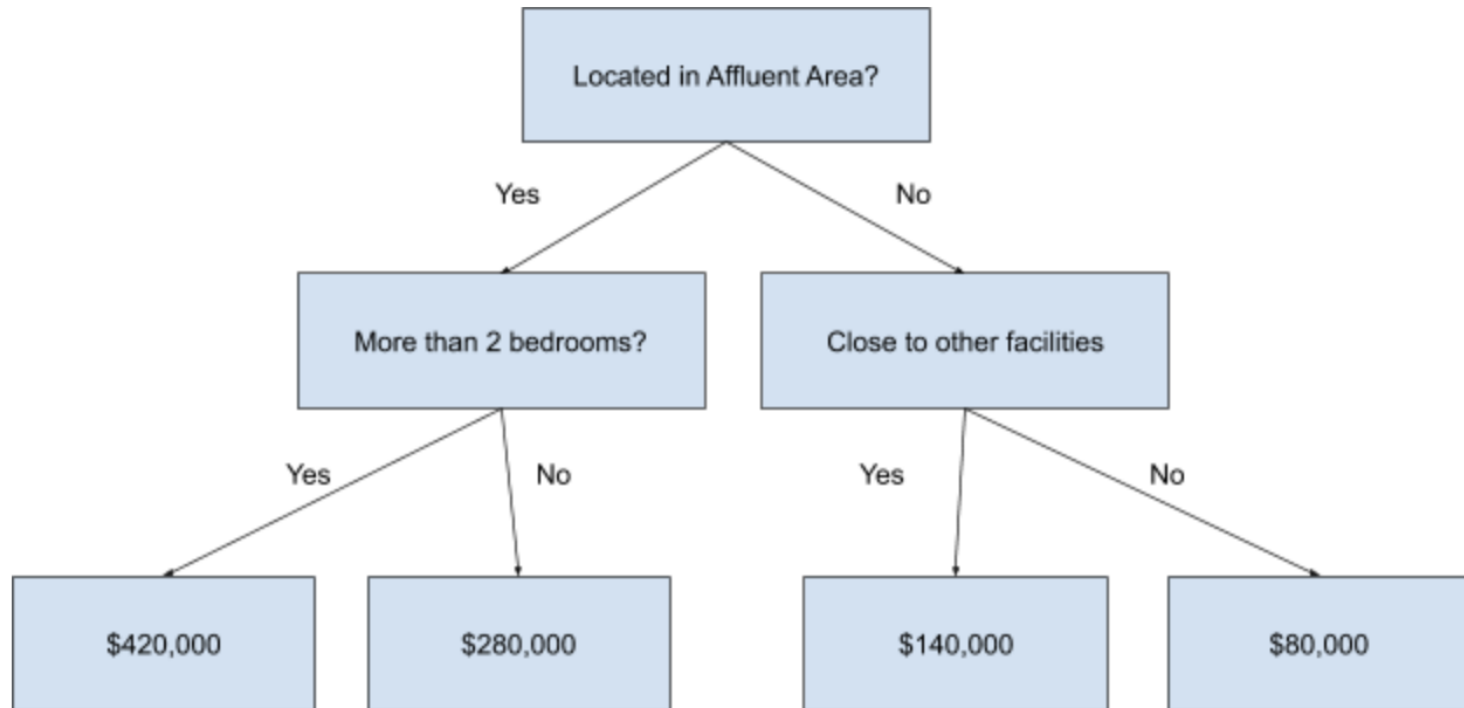
Decision Trees





How do you think?

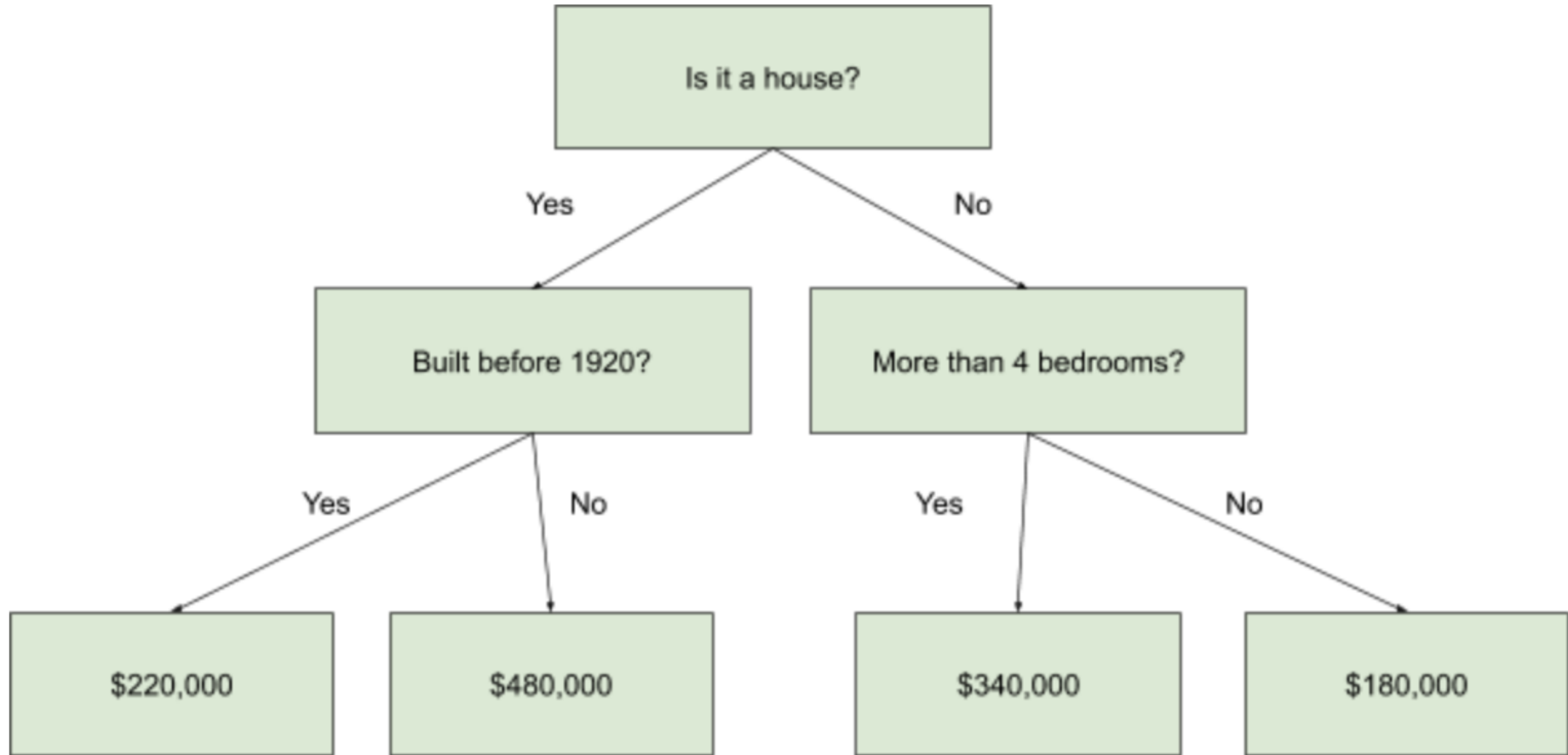
Buying a house: The obvious thing to do would be to look at historic prices of houses sold in the area, then create some kind of decision





How do you think?

—

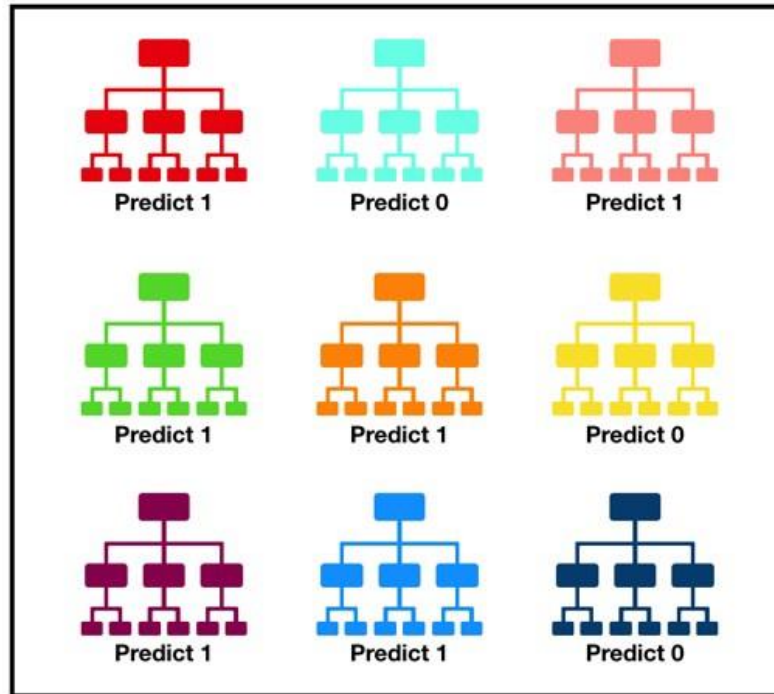




Random Forest

—

The random forest regression algorithm takes advantage of the 'wisdom of the crowds'. It takes multiple (but different) regression decision trees and makes them 'vote'



multiple different
decision
trees

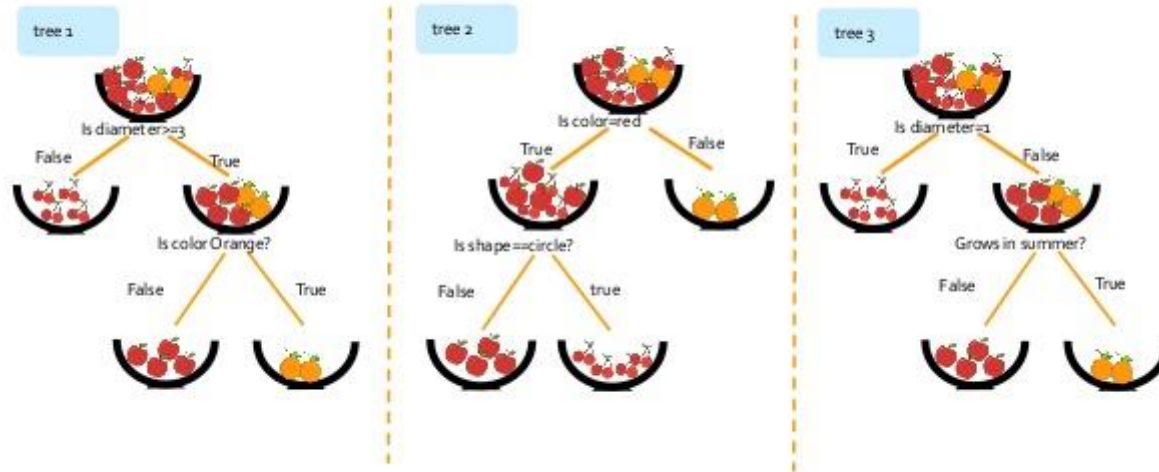




Random Forest

Each tree needs to predict the expected price of the real estate based on the decision criteria it picked

How does a Random Forest work?



simplilearn

©Simplilearn. All rights reserved.

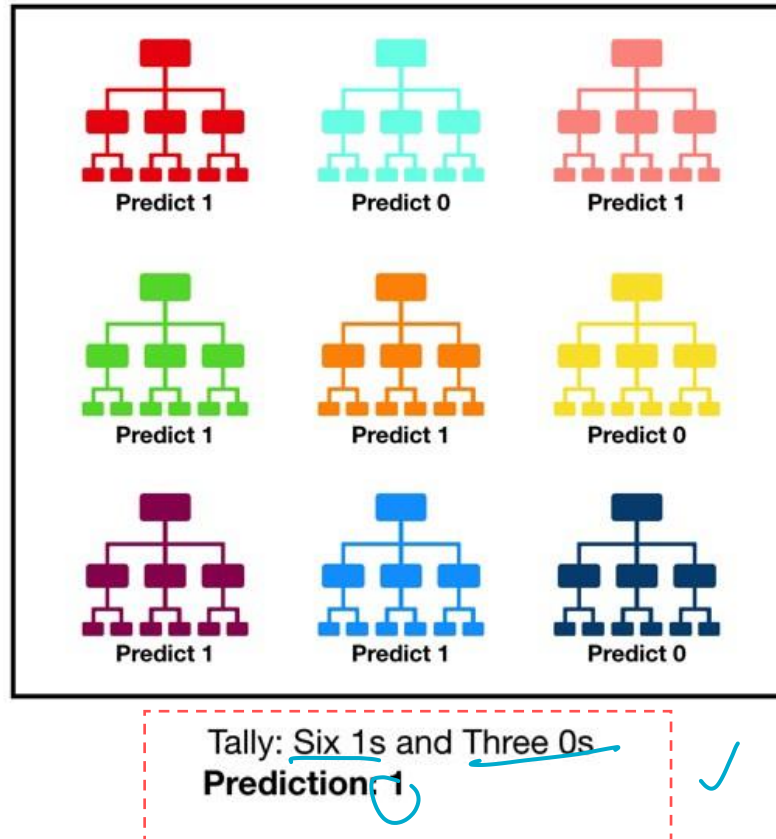
ใช้แบบสุ่ม
การสุ่ม



INVESTIC



Random Forest





Random Forest

Extremely high accuracy

It usually produces better results than other linear models, including [linear regression](#) and [logistic regression](#).

• Random Forest → ensemble over fit (good generalize than)

Scales well

Computationally, the algorithm scales well when new features or samples are added to the dataset.

scale 10000

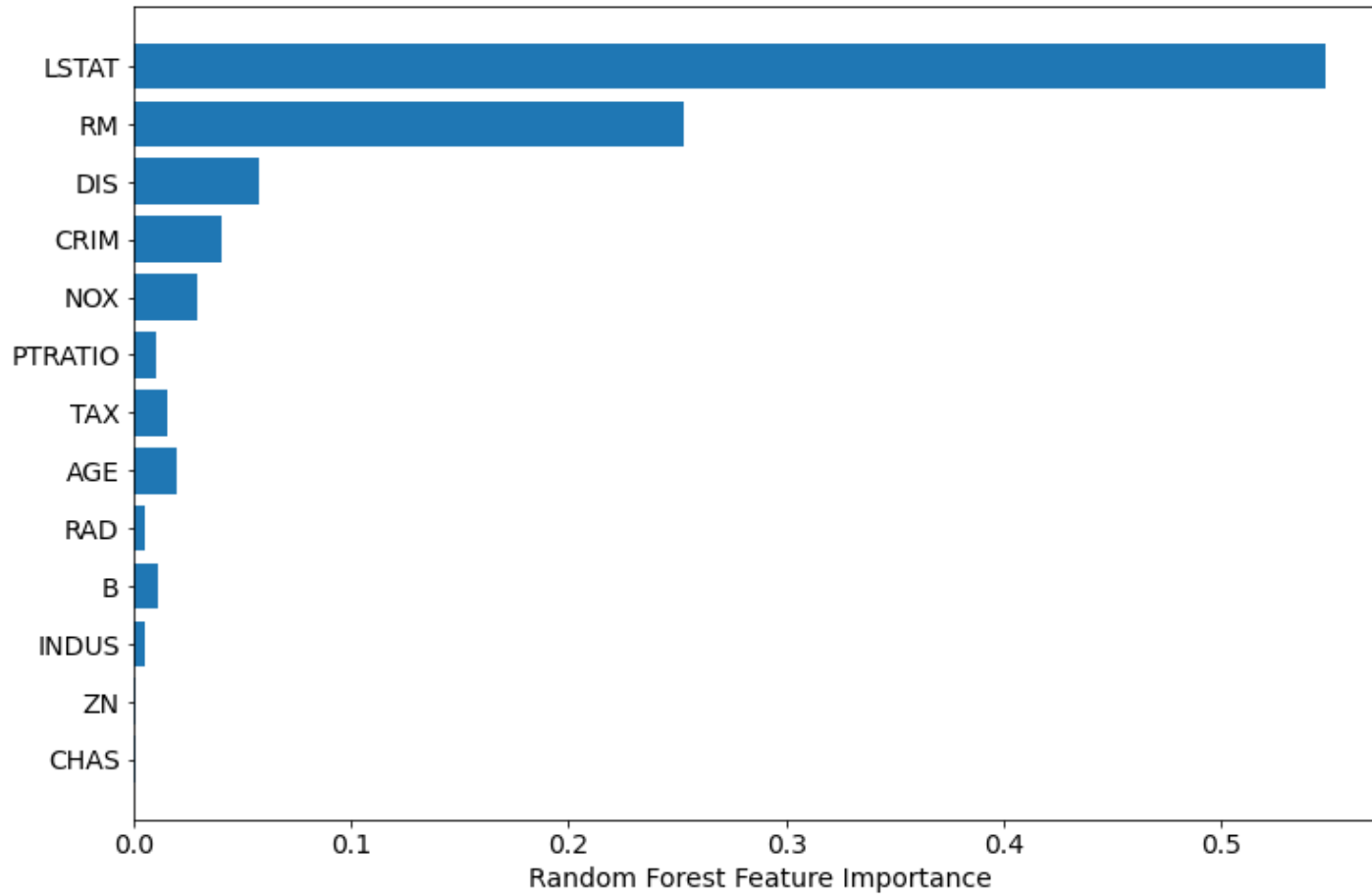
Interpretable → 0.8 accuracy (0.7 accuracy 1/10)





Feature Importance

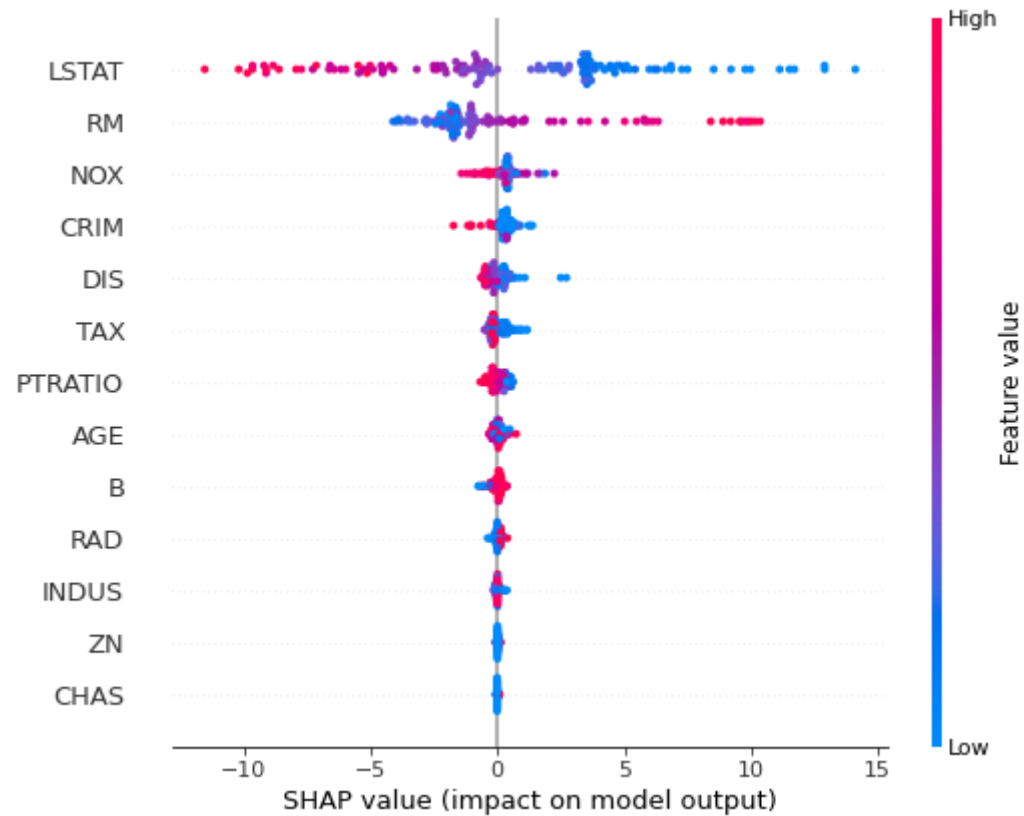
Interpretable





Feature Importance

Interpretable





Consider

A game – random 0-100 , if number >40 = win ✓

1. **Game 1** — play 100 times, betting \$1 each time.
2. **Game 2** — play 10 times, betting \$10 each time.
3. **Game 3** — play one time, betting \$100.

*Expected Value Game 1 = $(0.60 * 1 + 0.40 * -1) * 100 = 20$*

*Expected Value Game 2 = $(0.60 * 10 + 0.40 * -10) * 10 = 20$*

*Expected Value Game 3 = $0.60 * 100 + 0.40 * -100 = 20$*

1/10 2/10 3/10 4/10 5/10 6/10 7/10 8/10 9/10 10/10

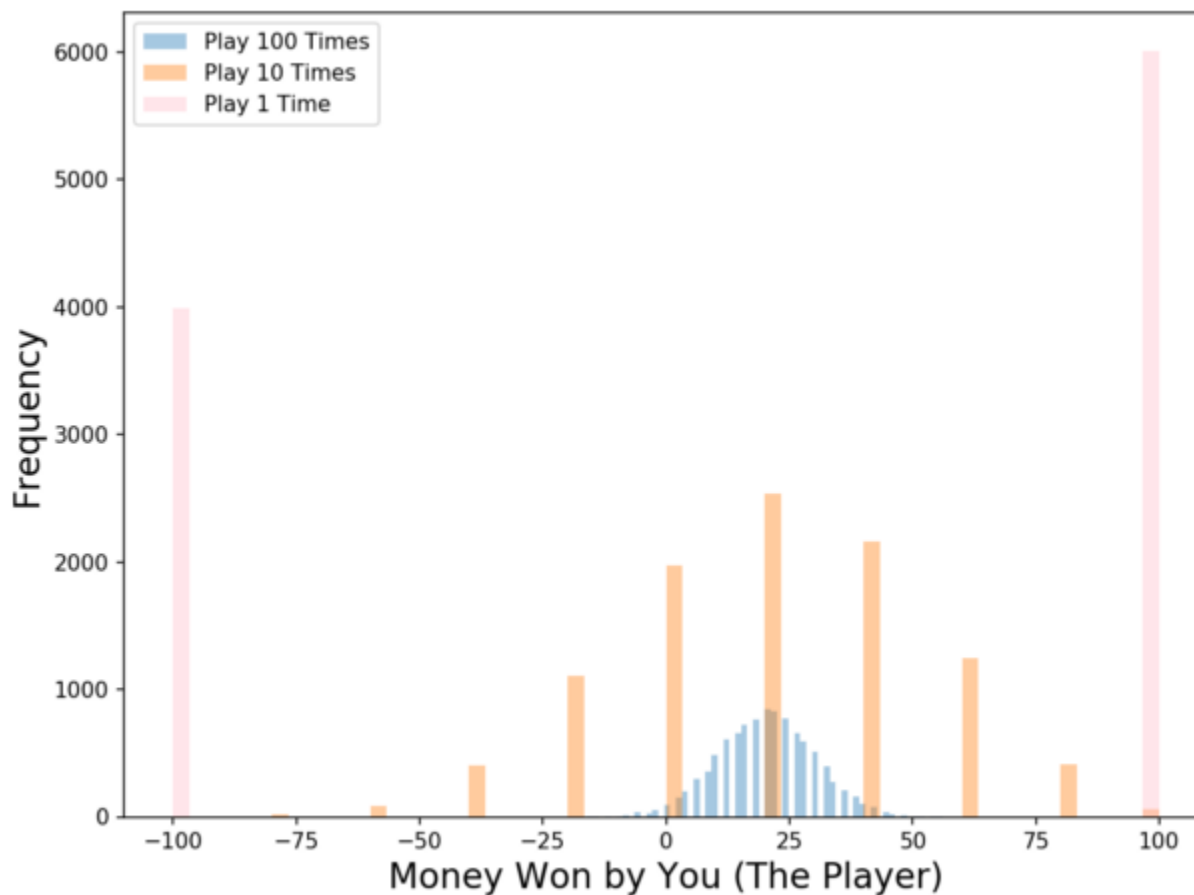




Monte Carlo Simulation

→ 9500€

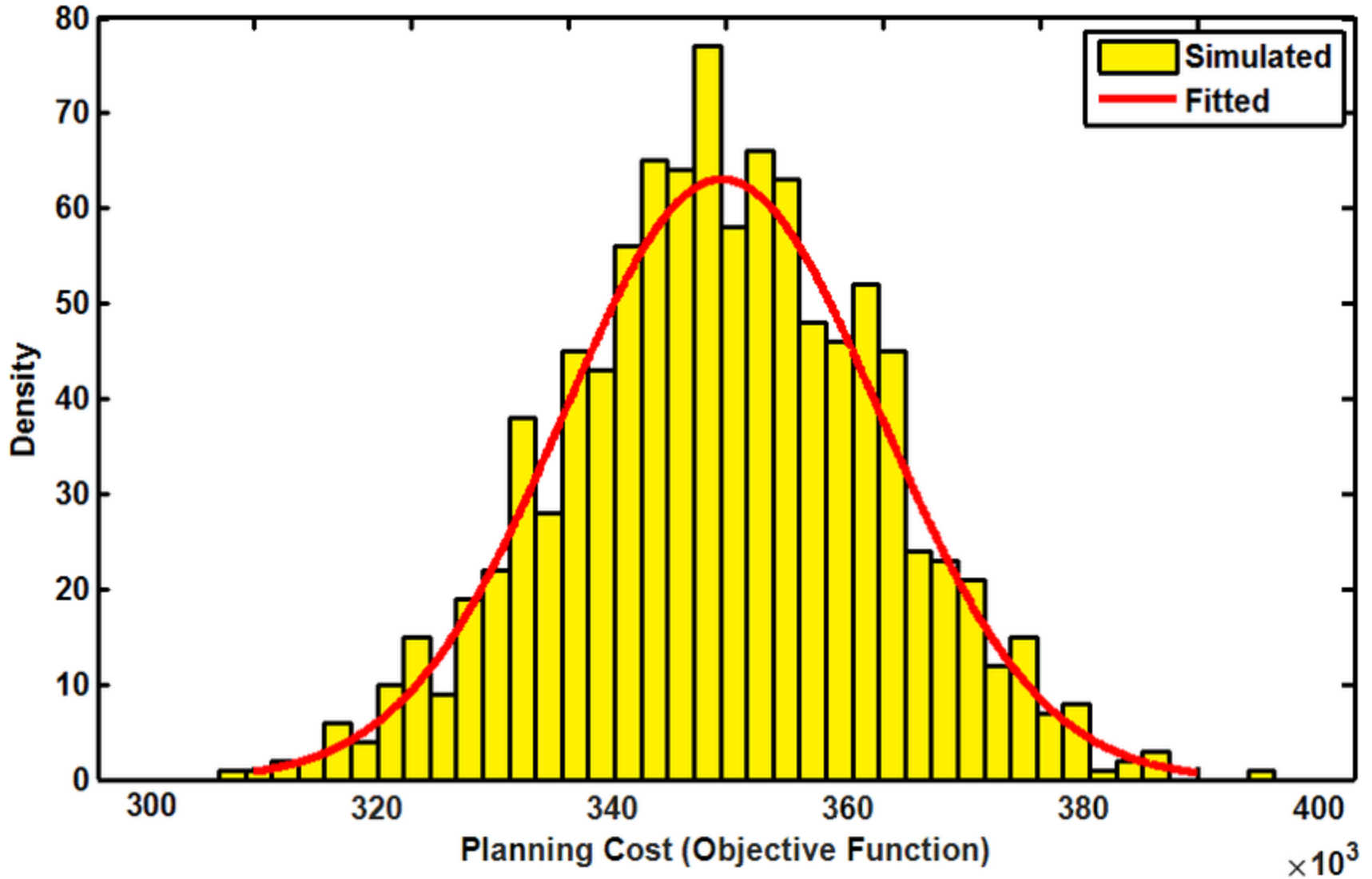
bell
curve



INVESTIC

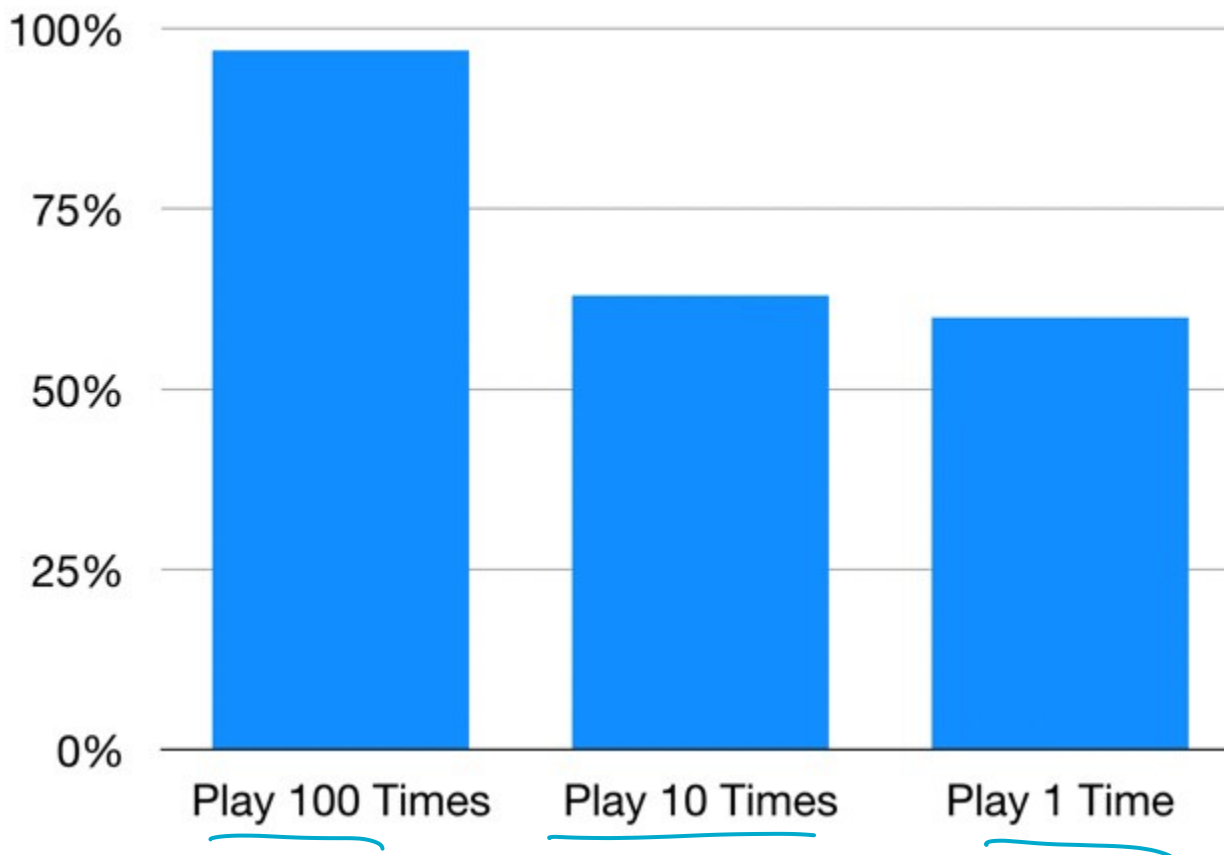


Monte Carlo Simulation





Monte Carlo Simulation



Random forest is the same — each tree is like one play in our game earlier.



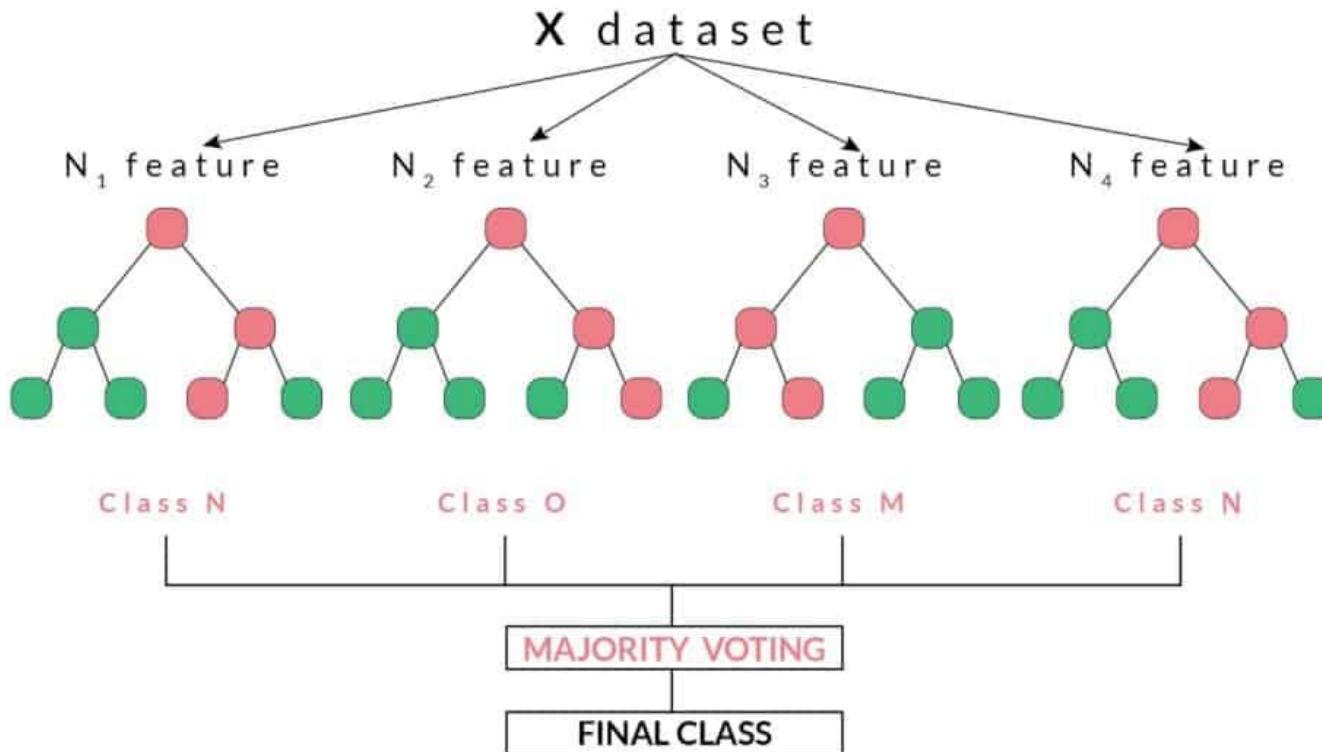
INVESTIC

הכל: 3000 משחקים
3000 Games



Random Forest

OVERFIT?





Understand Model

Random forest is both a **supervised learning** algorithm and an **ensemble** algorithm.

← 522 *WOLN Disiein Tree*

Ensemble algorithms **combine** multiple other machine learning algorithms,

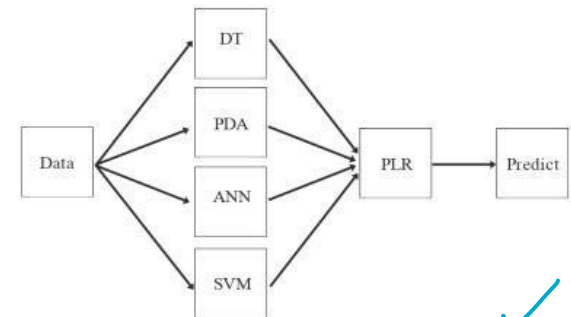
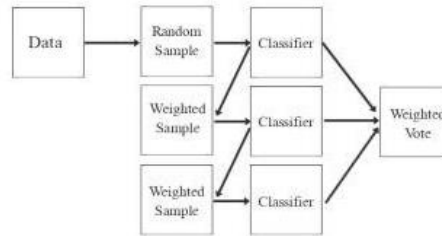
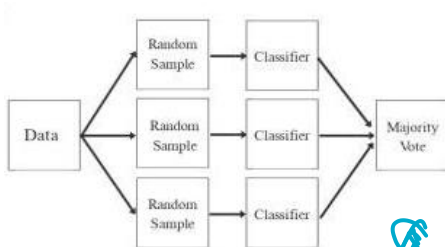
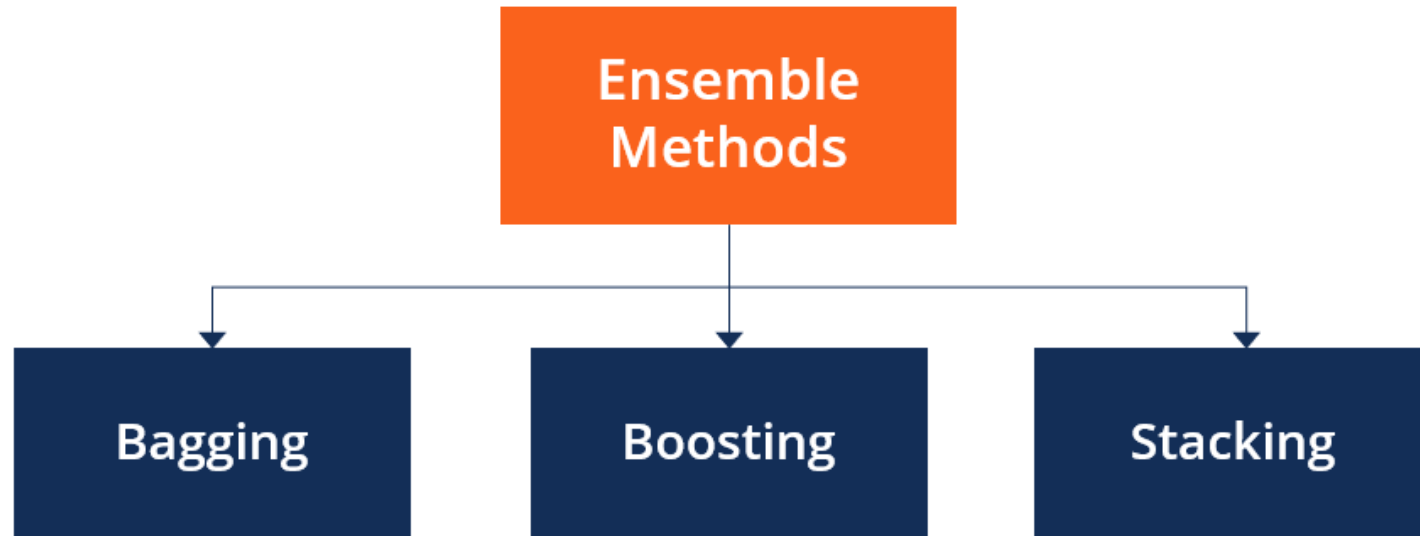
in order to make more accurate predictions than any underlying algorithm could on its own.

In the case of random forest, it ensembles multiple decision trees into its final decision.





Understand Model



Random forest is
the Bagging

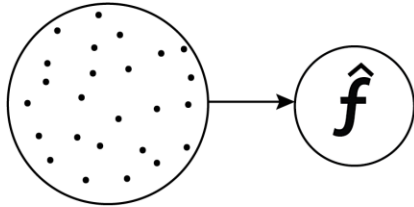




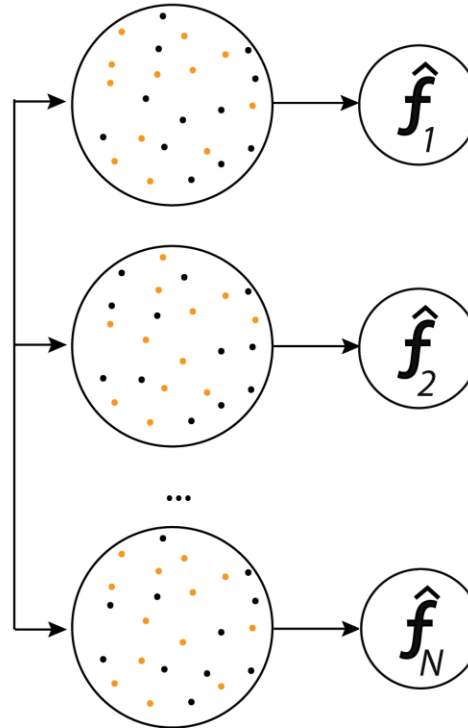
Bagging vs boosting

Handwritten note in blue: h_1, h_2, \dots, h_T
 \downarrow
Residuals

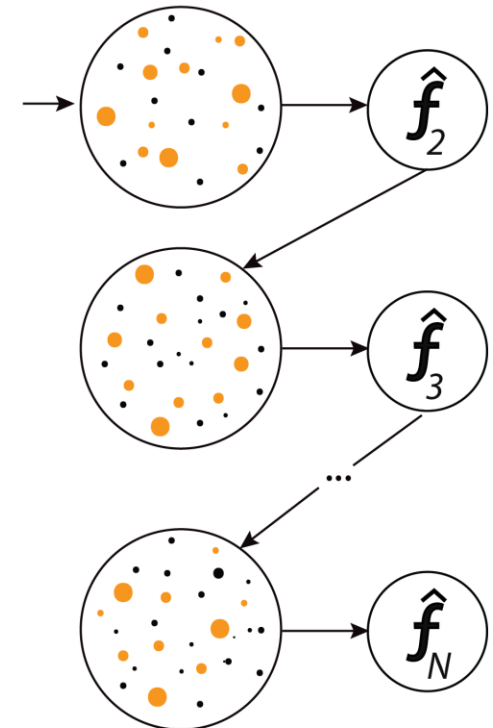
single



bagging



boosting

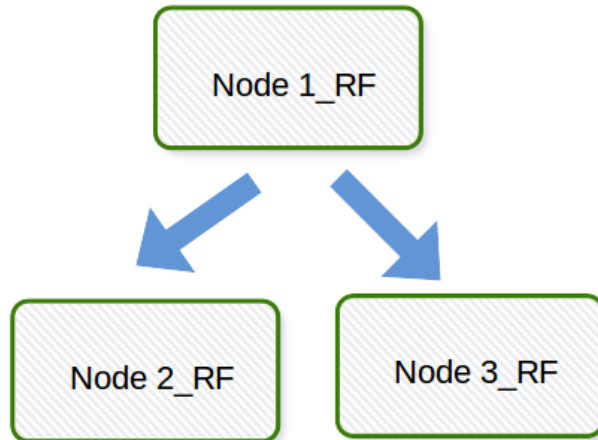




Random Forest vs Bagging Trees

Random forests--

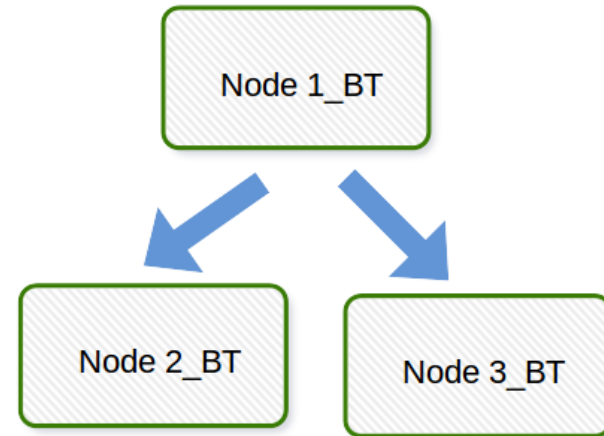
Only $m < M$ features considered
for each node for split



m can be selected via out-of-bag error,
but $m = \sqrt{M}$ is a good value to start with

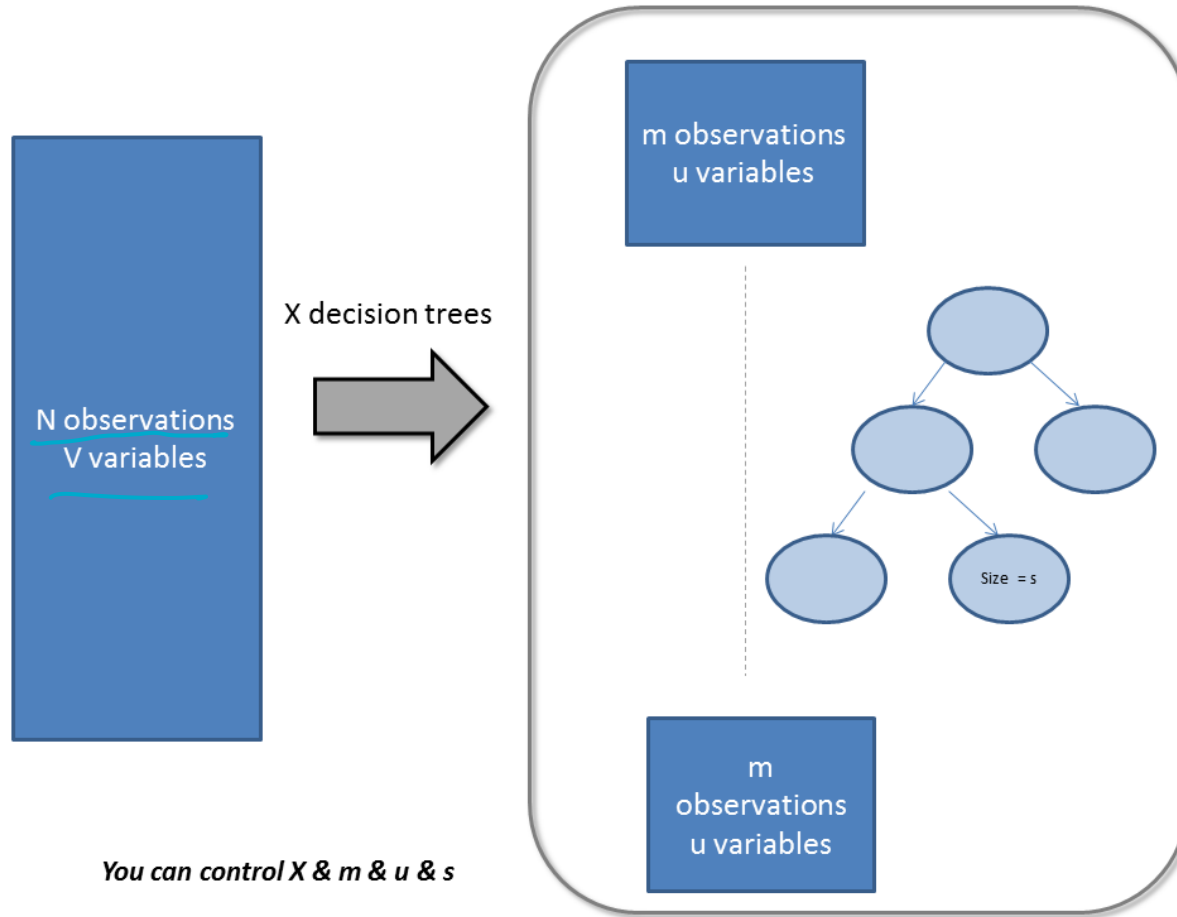
Bagging Trees--

All of M features considered
for each node for a split





Parameters ... play/optimize



See you soon at

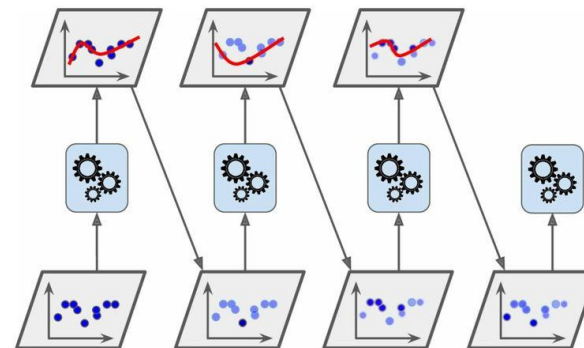
MACHINE LEARNING FOR INVESTING 101

2021

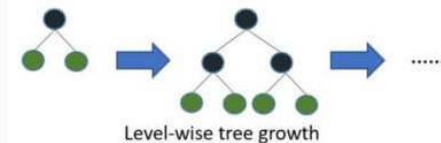


6400

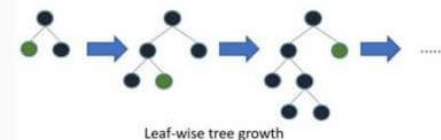
AdaBoost



XGBoost:



LightGBM:

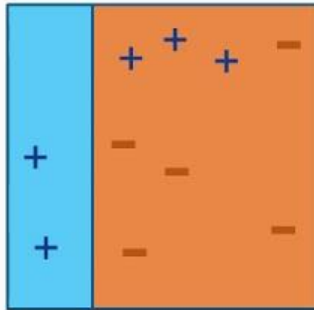




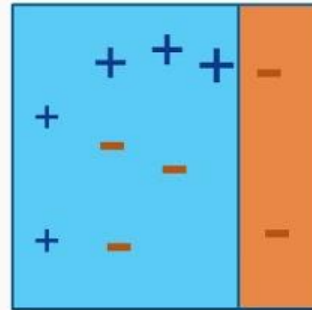
Boosting

Boosting

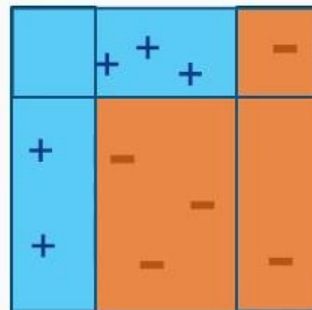
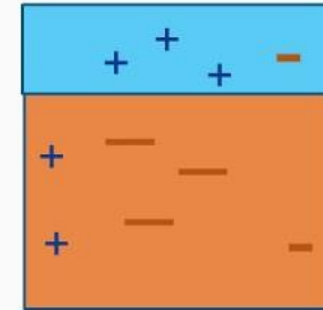
Iteration 1



Iteration 2



Iteration 3



Final Classifier/Strong classifier

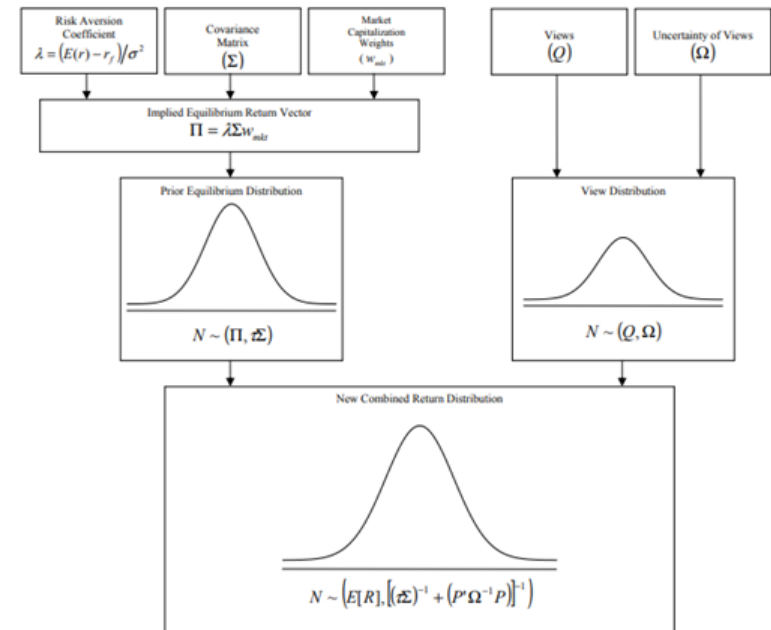
See you soon at

Coming Soon



PYTHON FOR
PORTFOLIO MANAGEMENT

ใช้ PYTHON ช่วยจัดพอร์ตลงทุน





Low Risk , High Return

