Target variable:

n = 1000

0 → 988

1 - 12

Discrete variable

(0, 1, 2...)

(0 - 1)

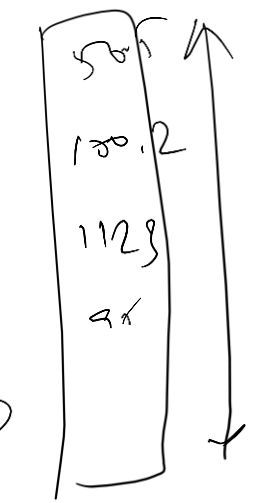
(Classification)

[Acc, CM, Class Report

Precision, Recall, F1-score]

Continuous variable (−∞ to ∞)

(Regression)



(Dis)

Classification

Logistic Regression

KNN classifier

NB classifier

D.T Classifier

RF "

(cont)

Regression

Linear Reg

KNN Regressor

D.T Regressor

RF "

(-∞ to ∞)

⇓

(0 to 1)

⇓

(0-1)

Log Reg

Bagging → Bootstrap Aggregating

→ Random subset of features

→ " " samples

→ " "

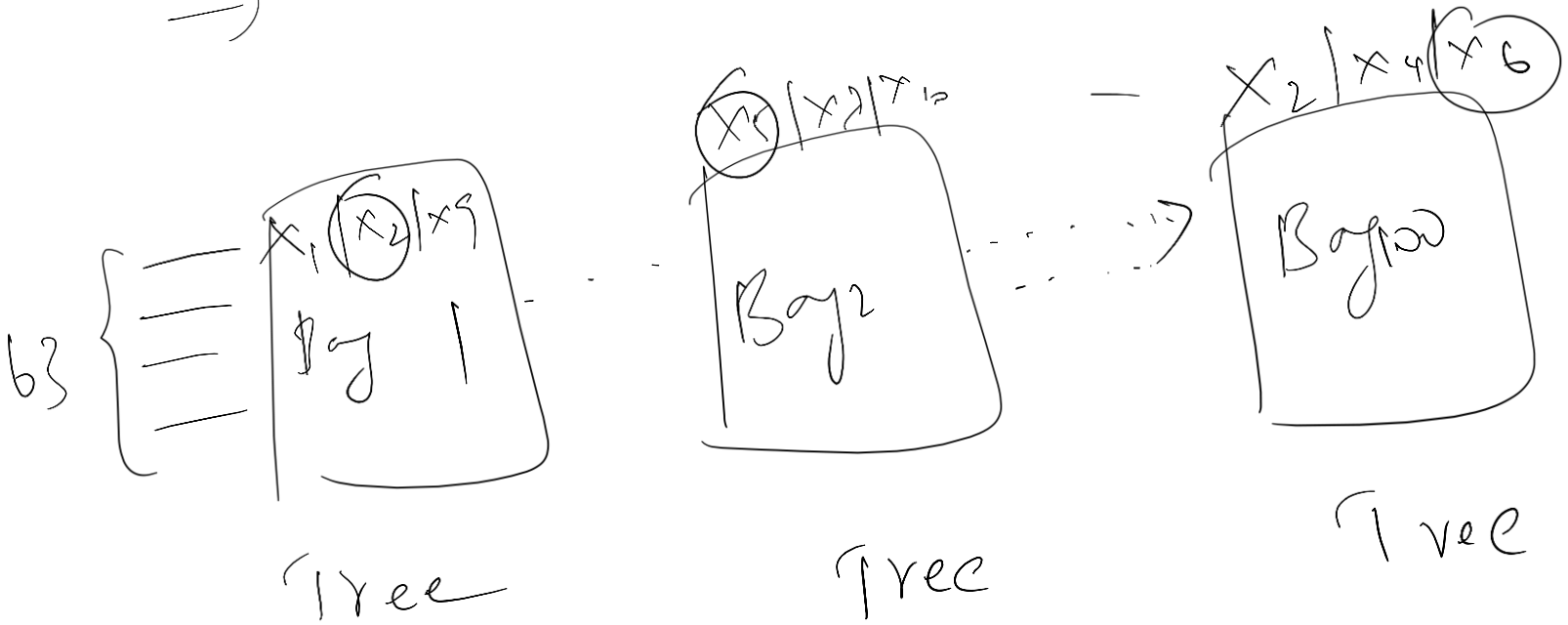
n = 100

(Training)

63%

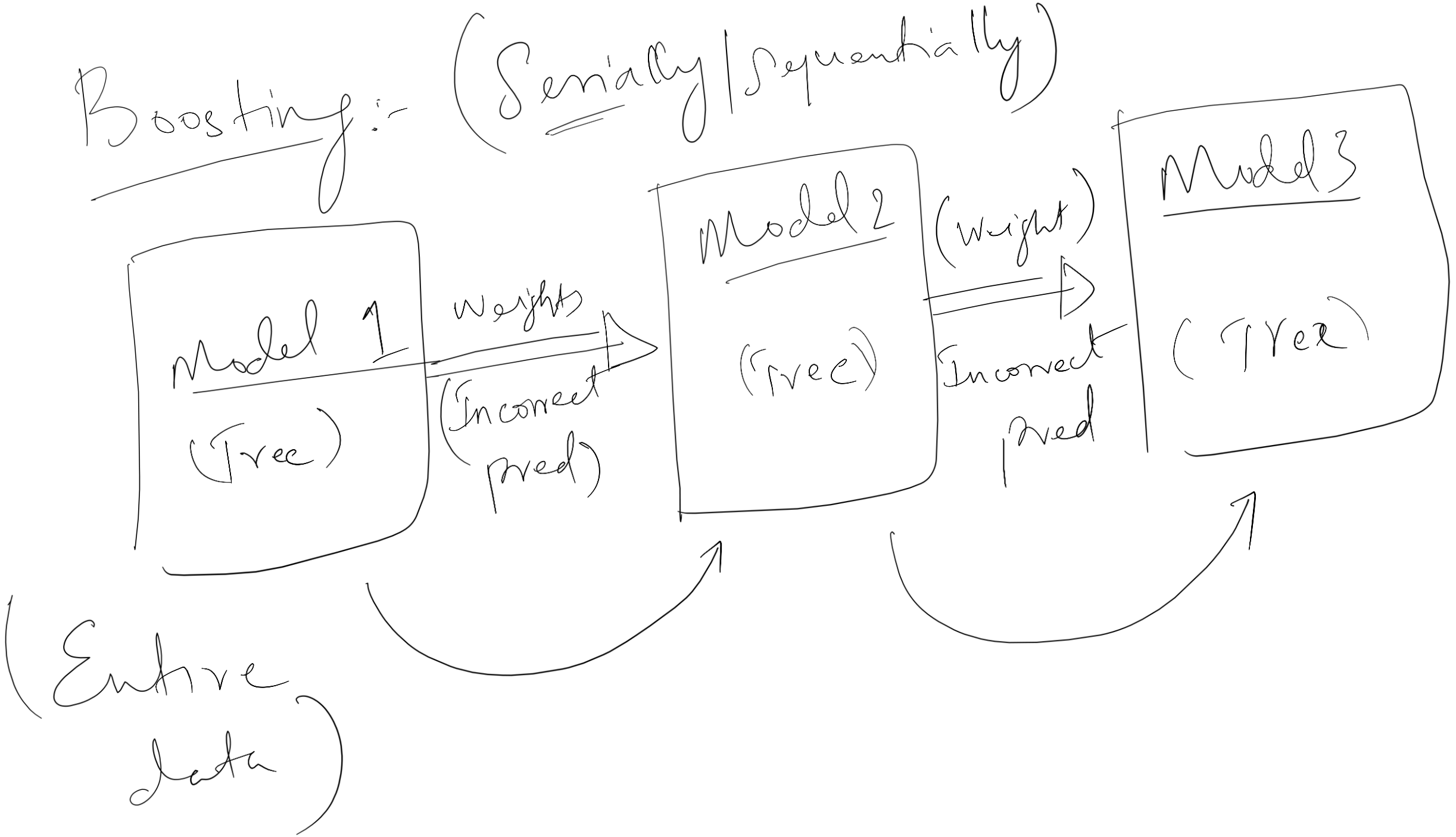
37%

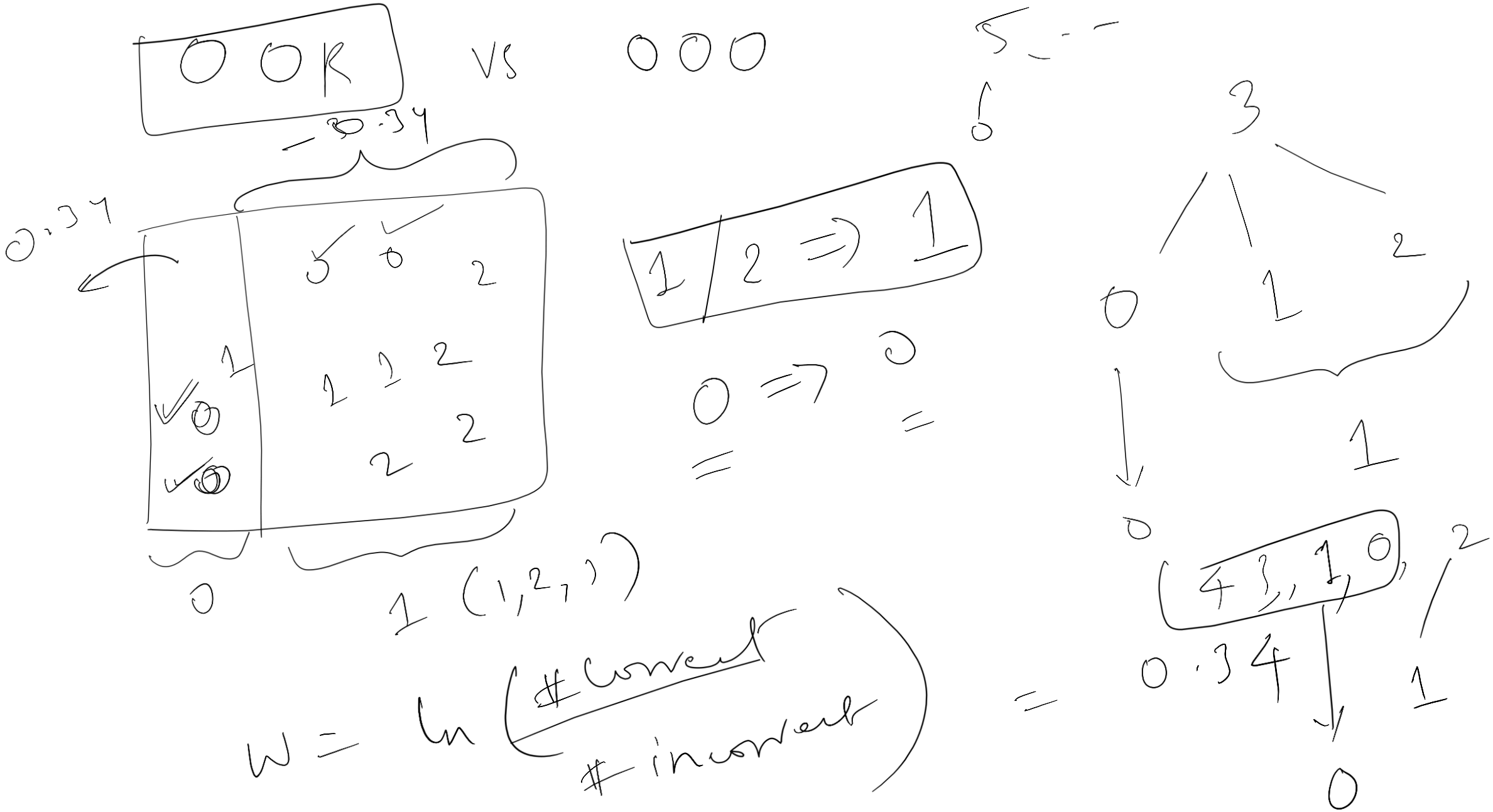
(oobscore)

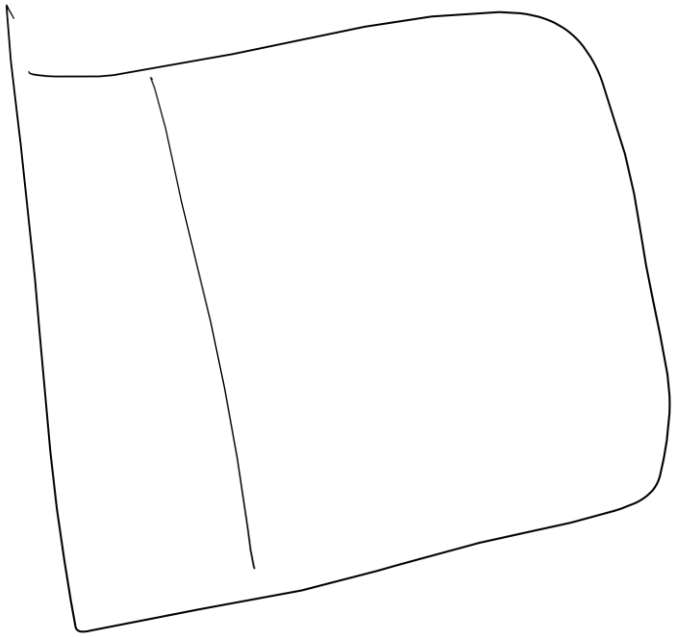


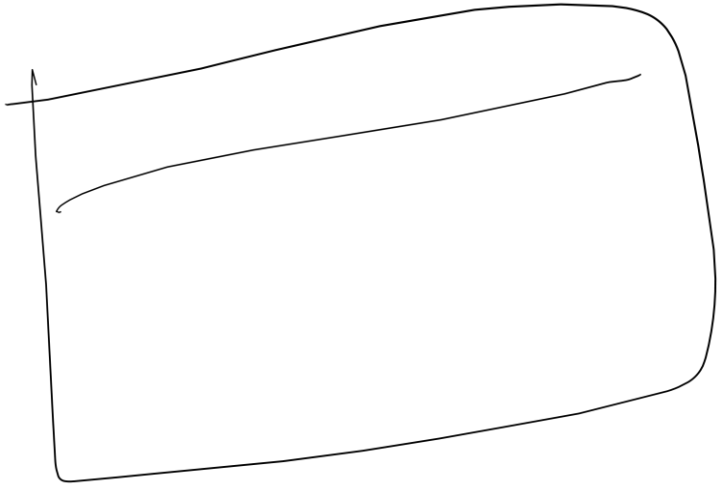
n = 10

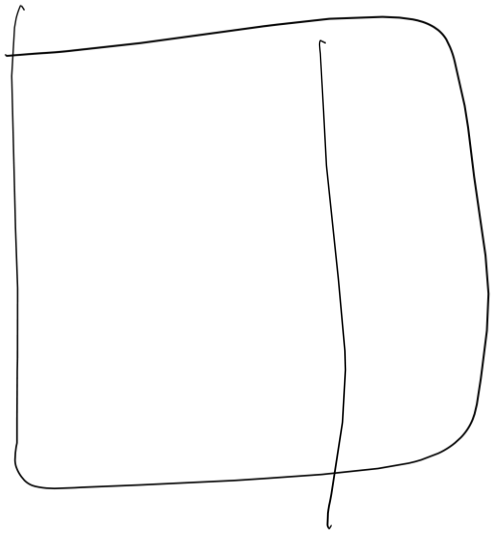
3

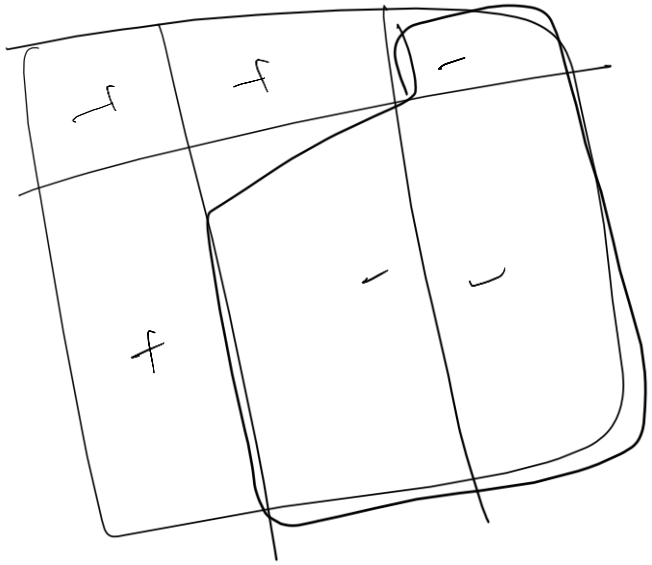

Boosting :-
(Serially | Sequentially)
Model 1
(Tree)
Weights
(Incorrect pred)
Model 2
(Tree)
(Weight)
Incorrect pred
Model 3
(Tree)
(Entire data)

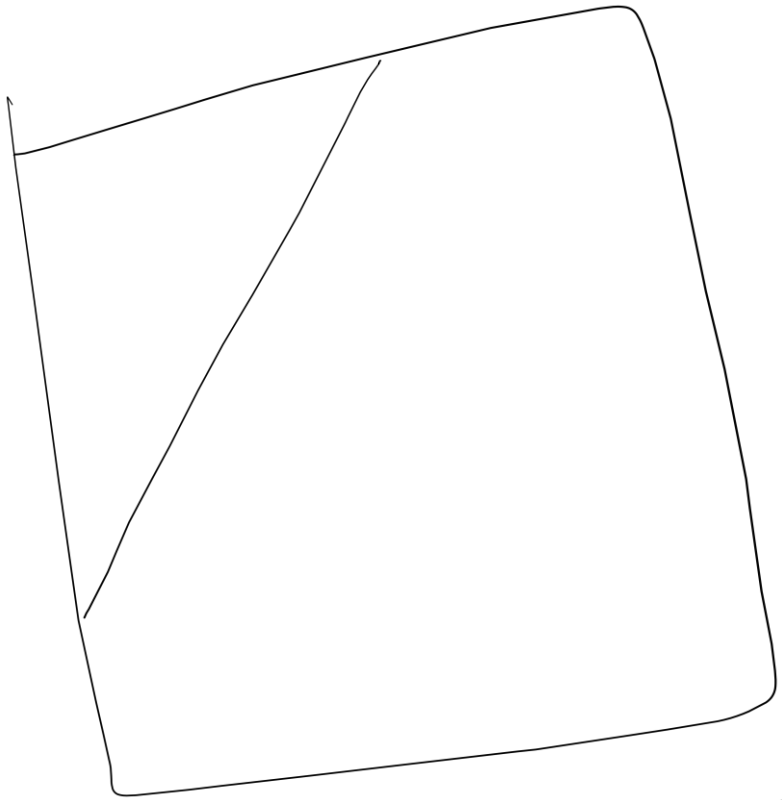

OOR
VS
000
-0.34
0.34
1
2
1
2
2
2
2
0
1 (1,2,1)
1
2 ⇒ 1
0 ⇒ 0
3
0
1
2
1
0
4, 3, 1, 0
0.34
1
0
2
W = ln (#correct / #incorrect)
=

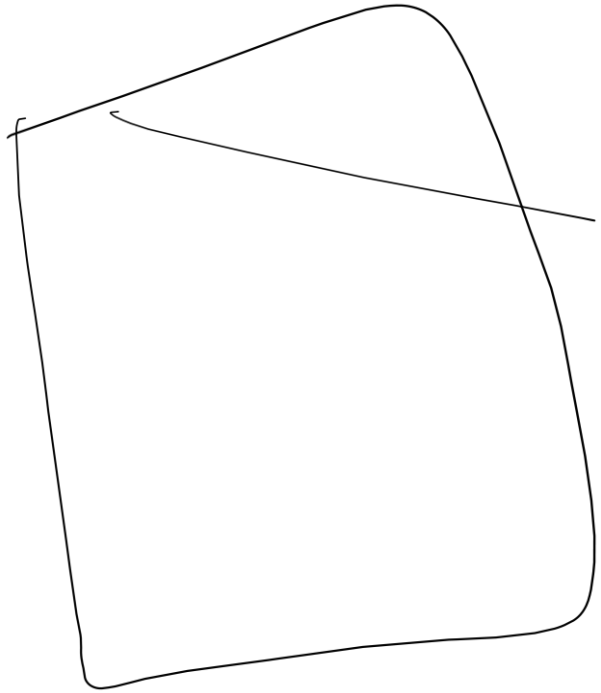


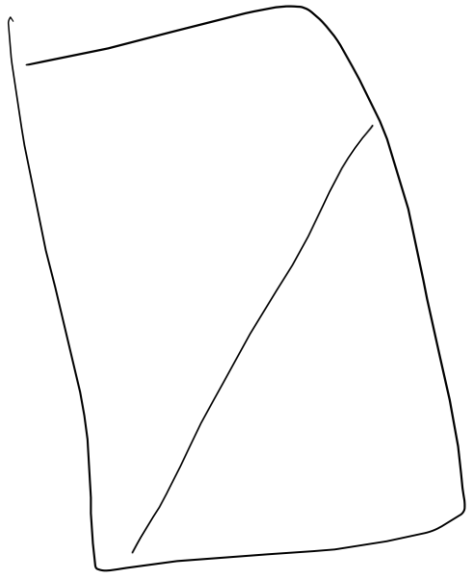


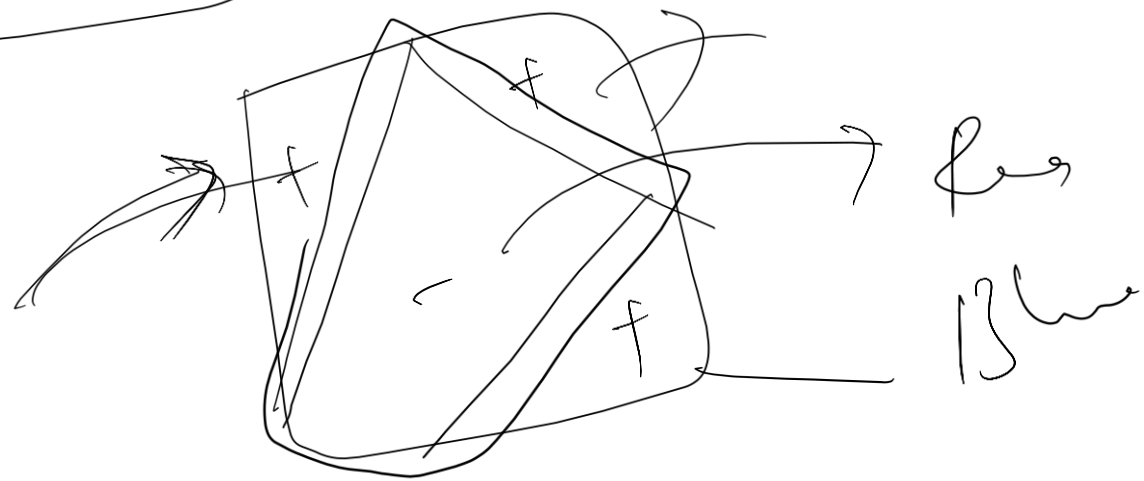


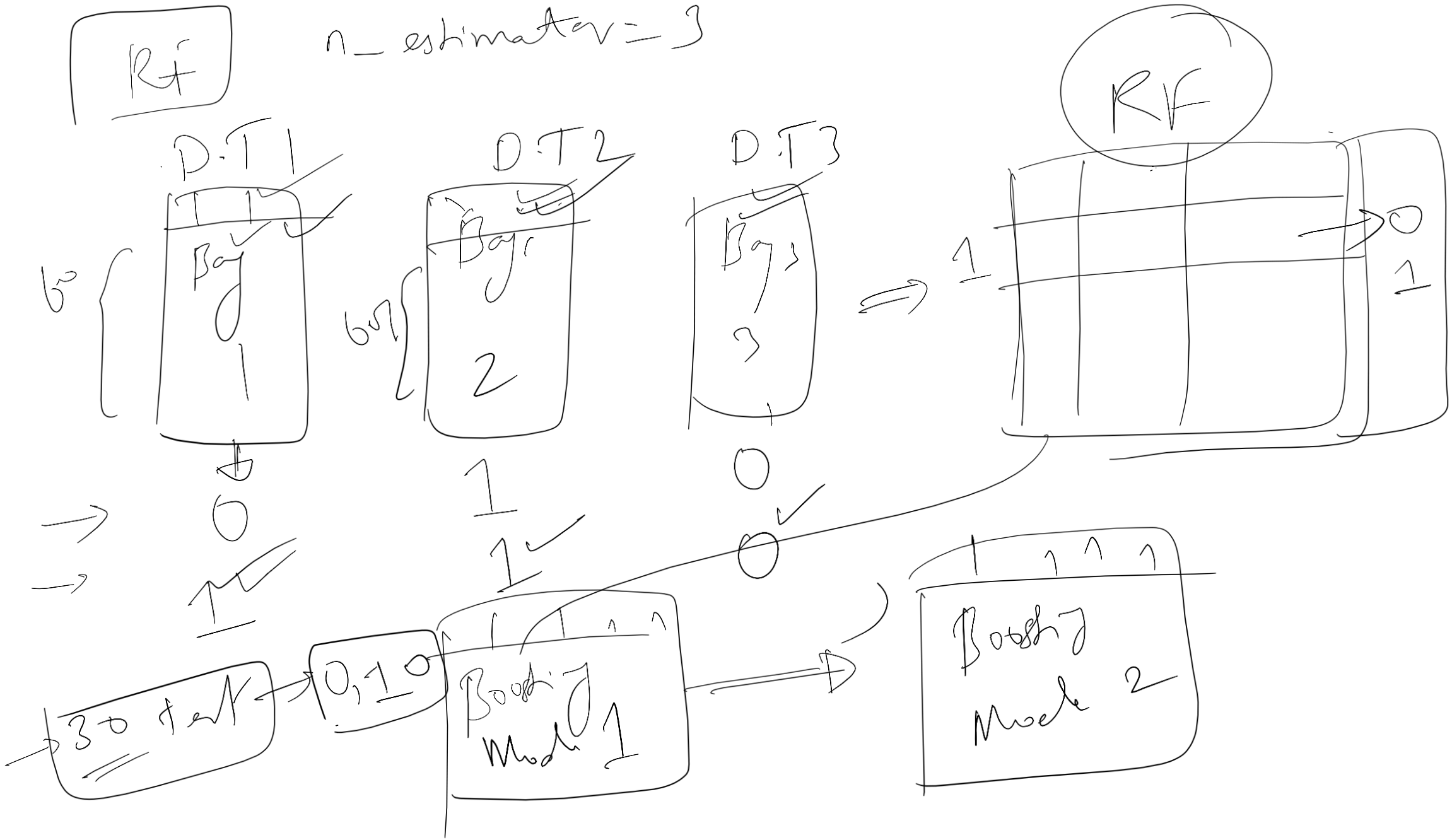


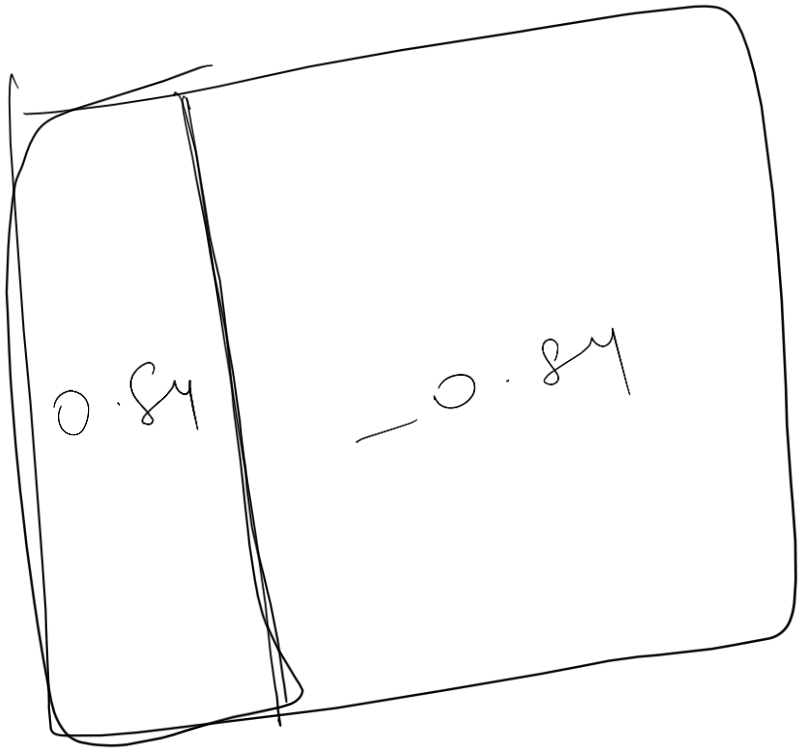


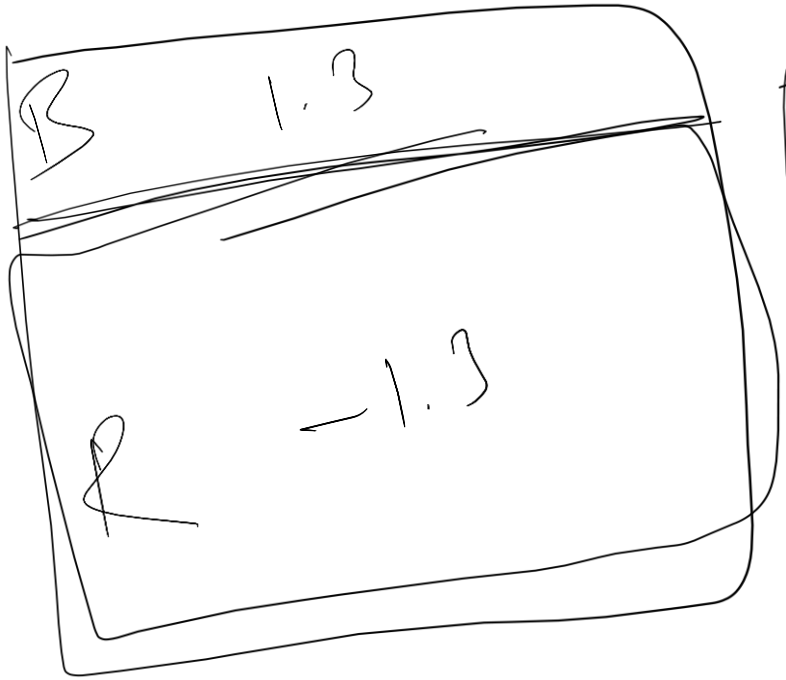


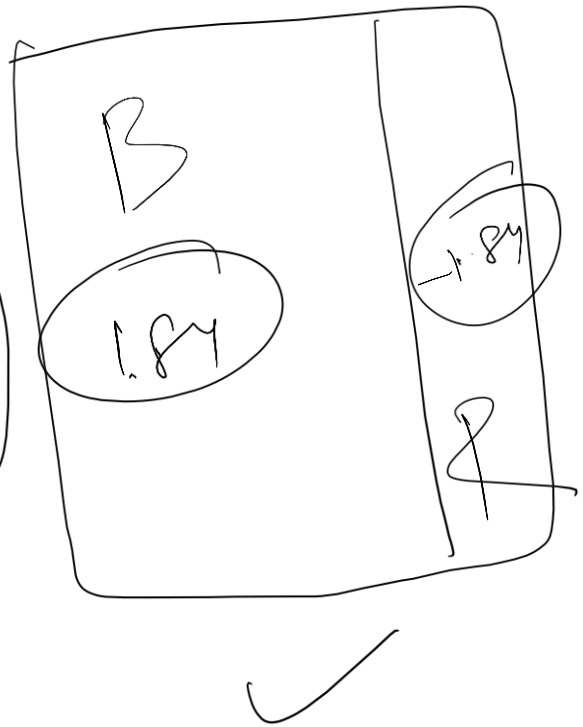



f
f
f
-

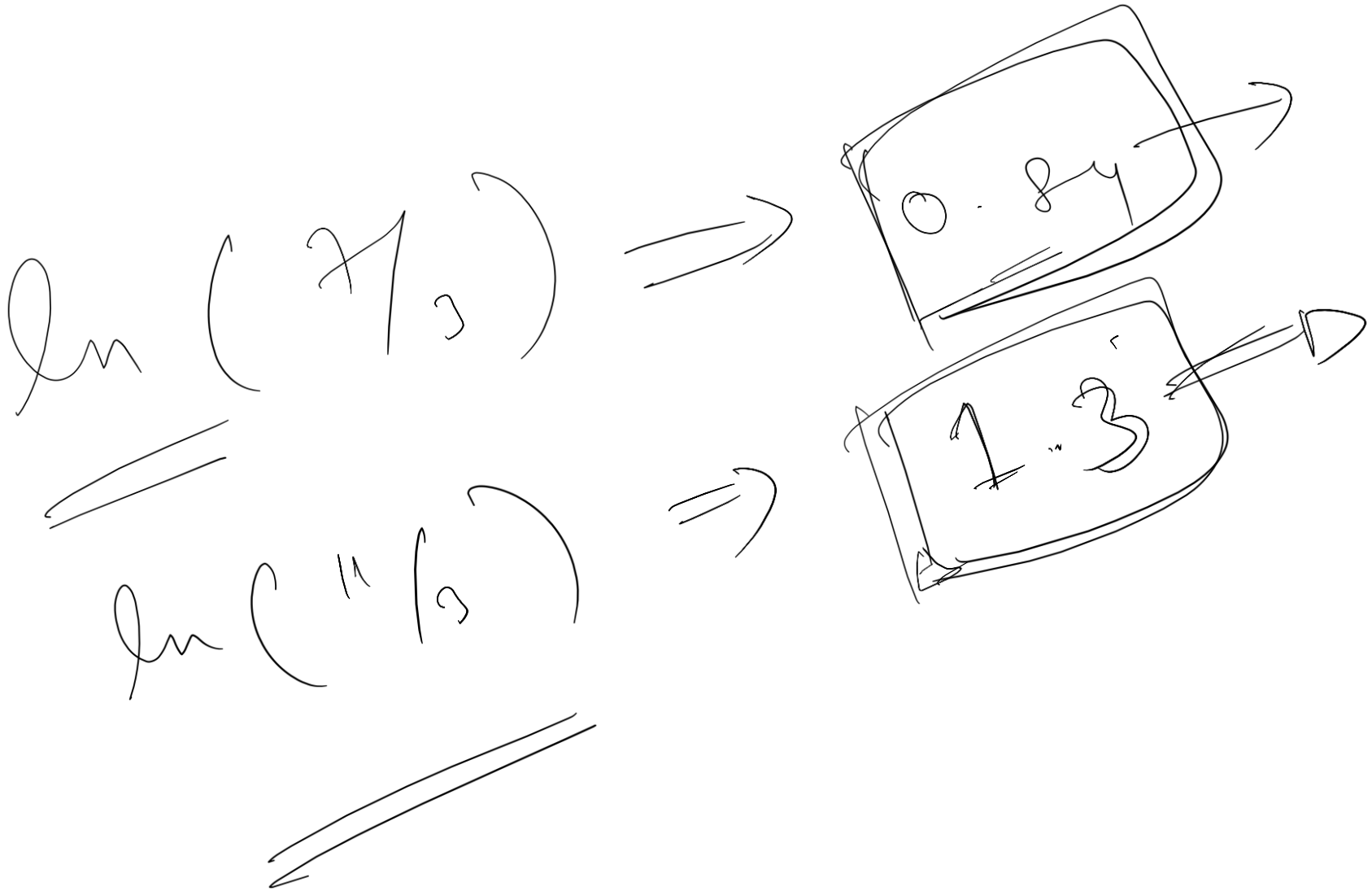

RF
n_estimators = 3
D.T 1
D.T 2
D.T 3
Bag
Bag
Bag
1
2
3
RF
1
1
0
1
1
0
0,1,0
Boosting
Model 1
Boosting
Model 2







1.84	1.84	−1.84
0.84 1.3	−0.84 1.3	−0.84 1.3
−1.3 0.84	−1.3 −0.84	−1.3 0.84



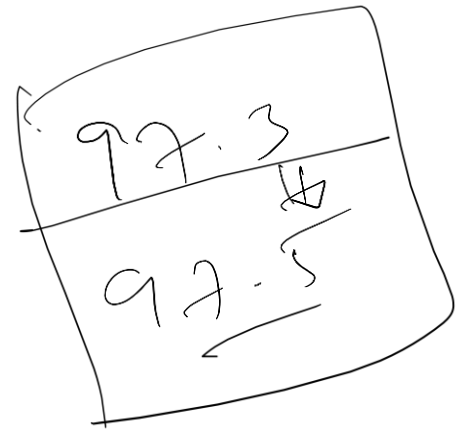
Voting: Classifier / Regressor

3 Models

1

2

3 AdaBoost



D.T

1 (5/6)

1

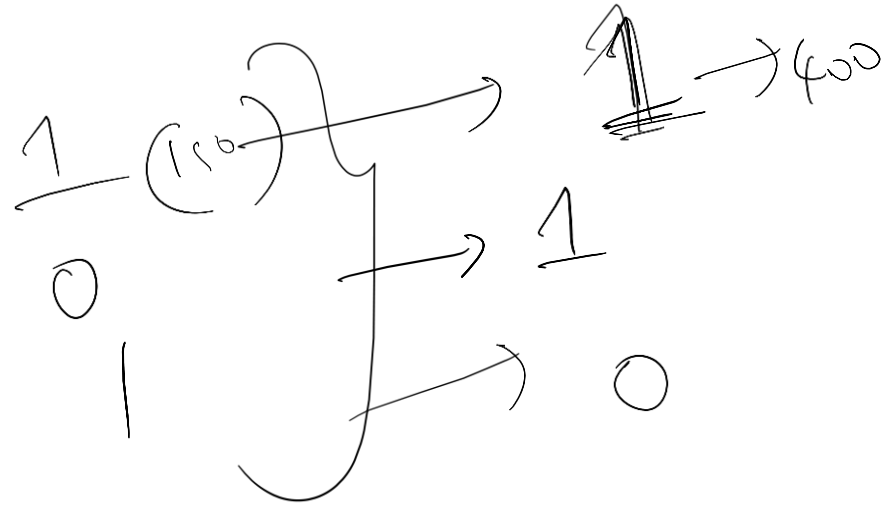
0

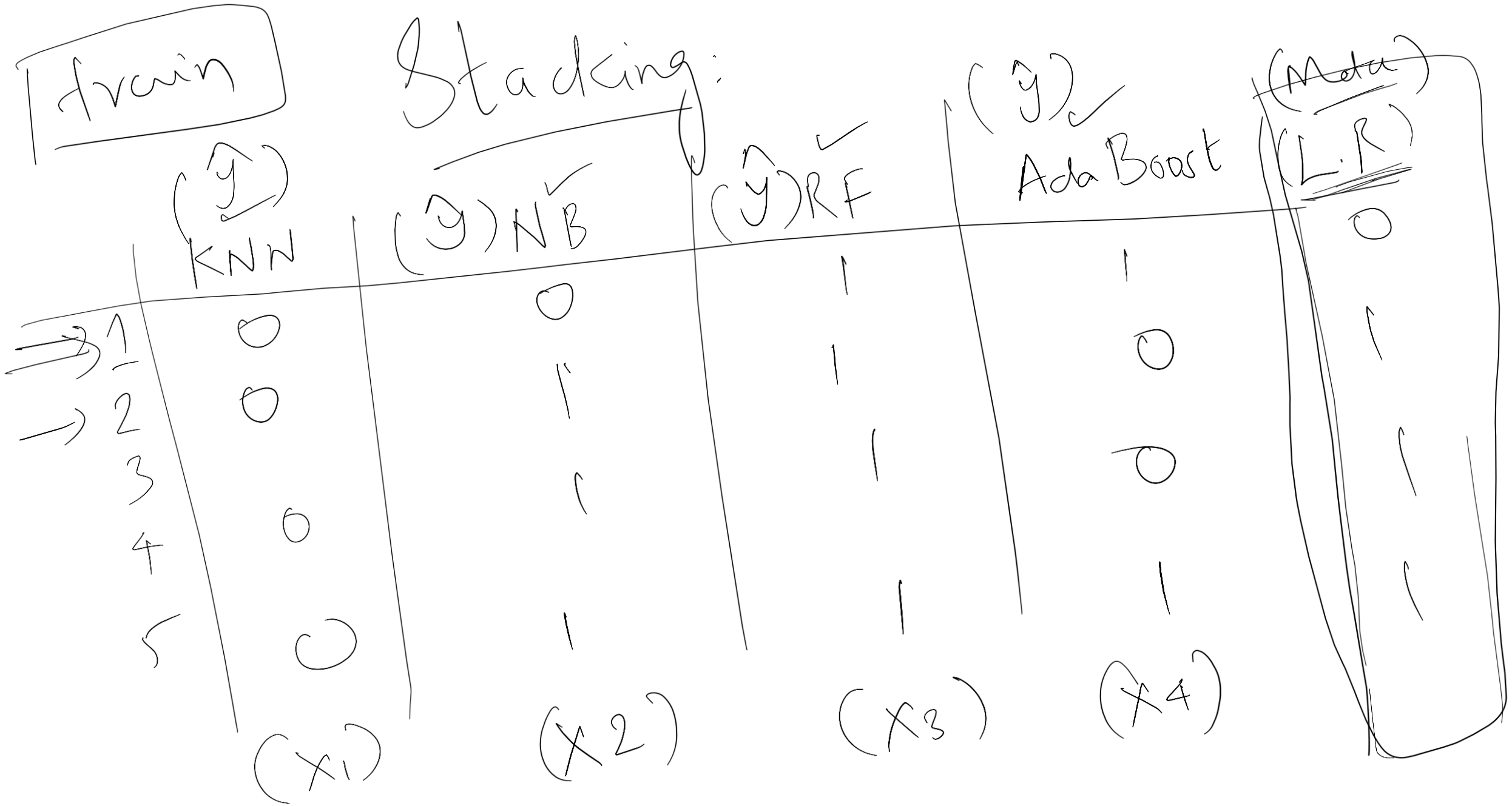
R.F

0 (3/6)

1

0

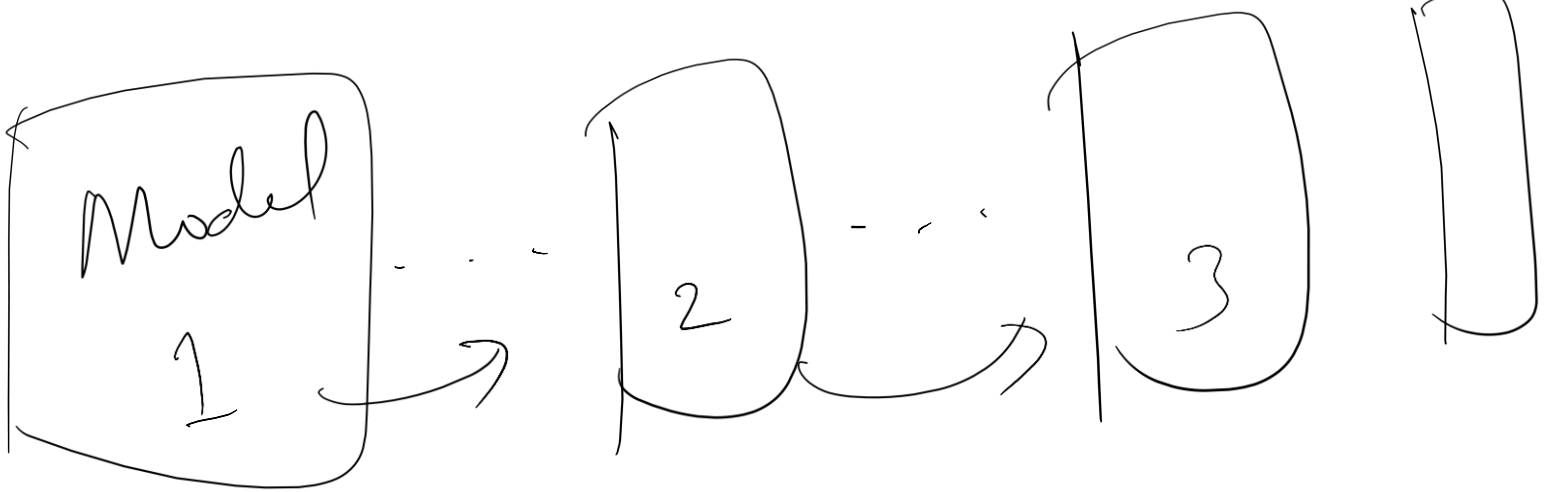



Train
Stacking:
(ŷ) KNN
(ŷ) NB
(ŷ) RF
(ŷ) Ada Boost
(Meta) (L.R)
1
2
3
4
5
0
0
0
1
1
1
1
1
1
1
0
1
0
0
1
0
1
1
1
(X1)
(X2)
(X3)
(X4)

Let's get back

at 7:50 P.M (I.S.T)

Ensemble :



Pred

Model

D.T

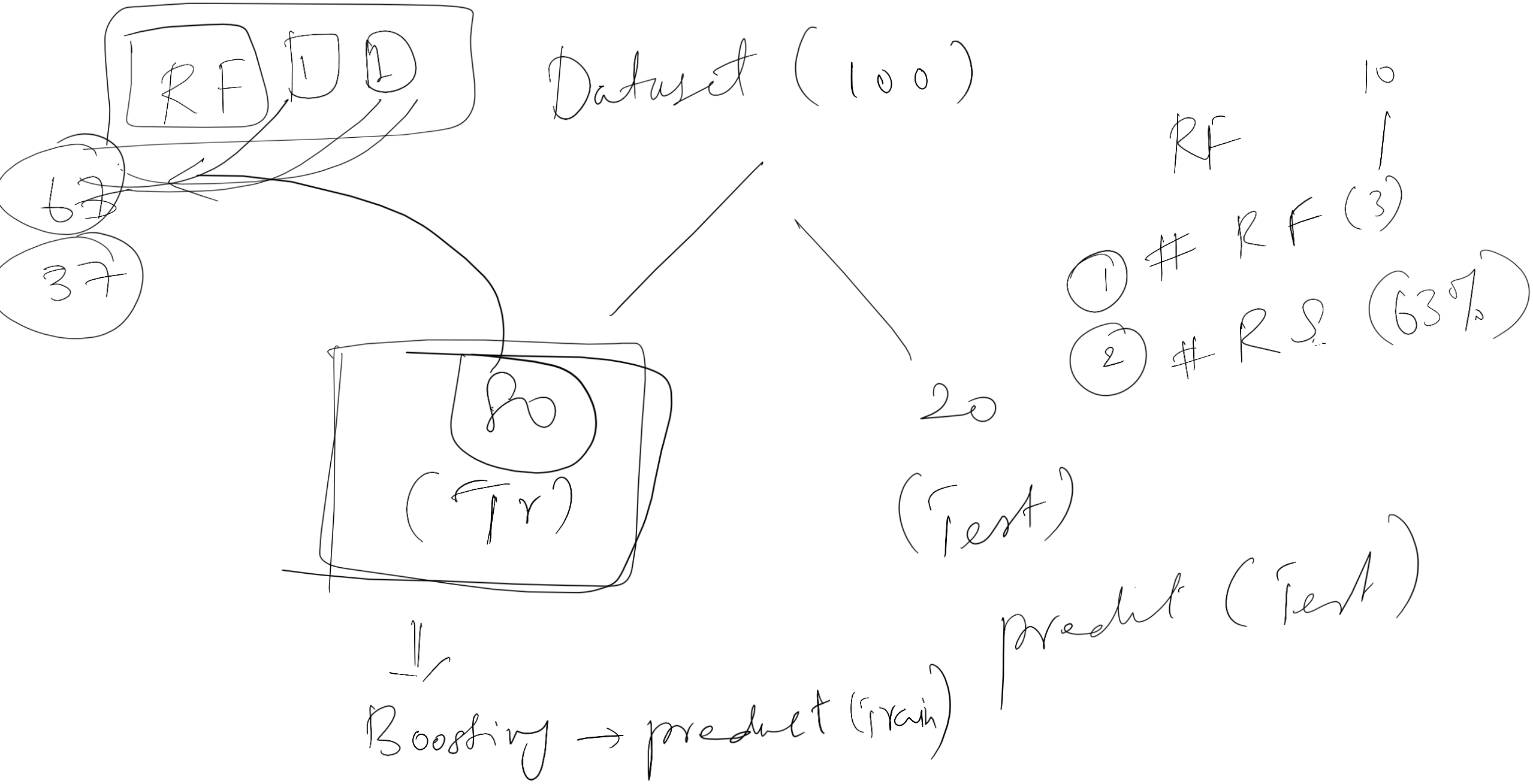
→ Bagging (RF)

→ Boosting (Ada, Xgboost ...)

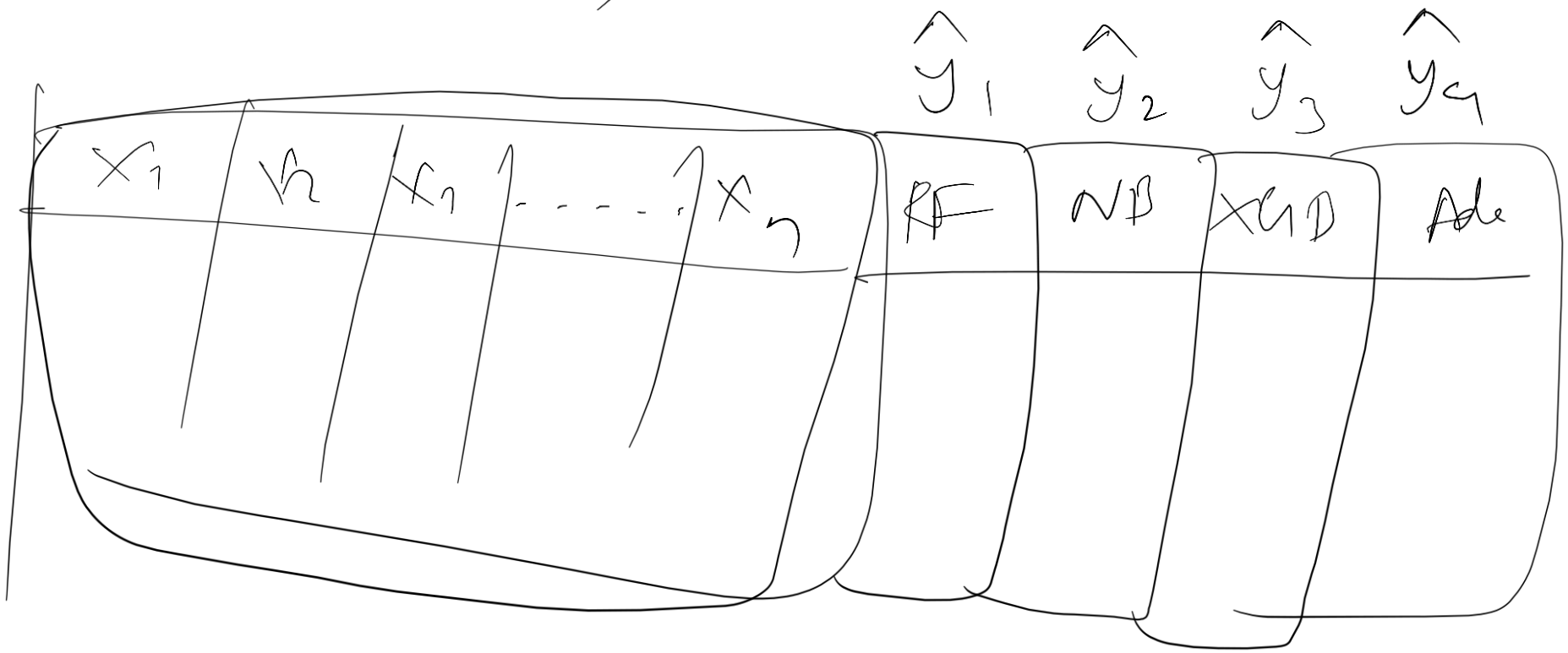
→ Voting → Votes of model pred

→ Stacking → Pred → I/p

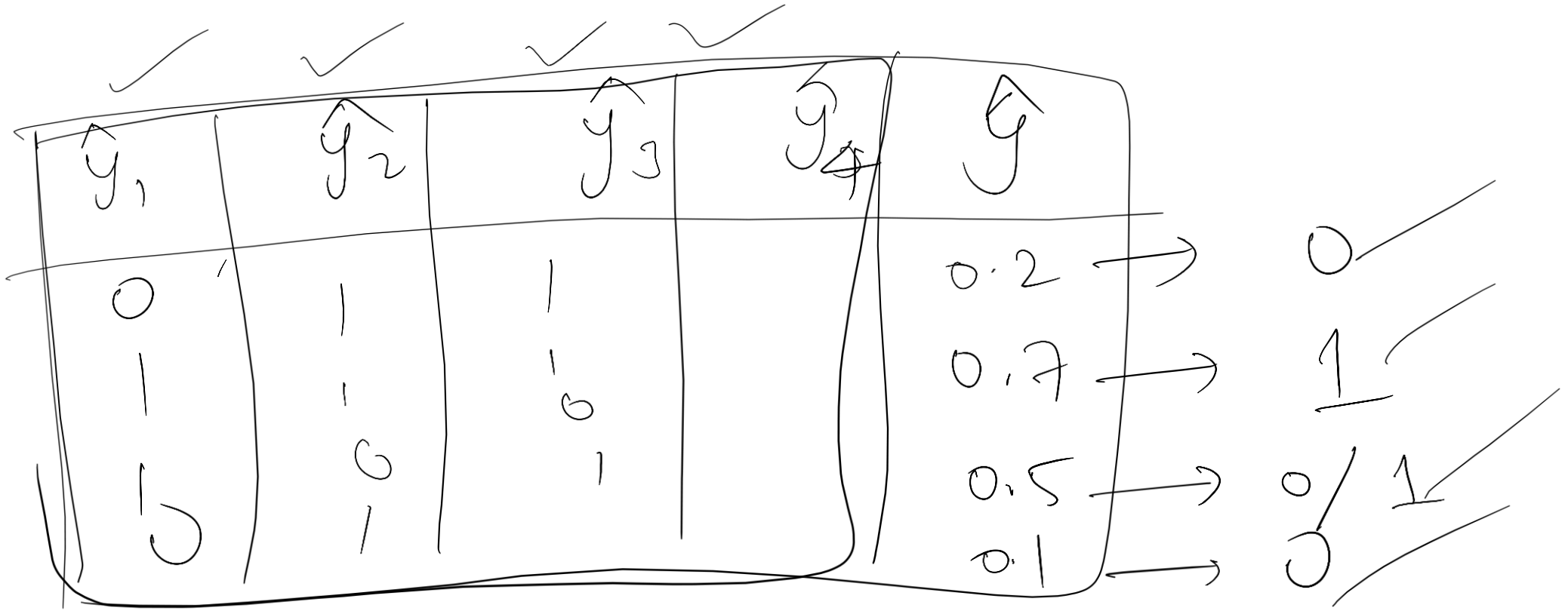
Meta $\xrightarrow{L.R}$ O/p


RF
1
1
63
37
80
(Tr)
Dataset (100)
20
(Test)
10
RF
1
RF (3)
1
RS (63%)
2
Boosting → predict (train)
predict (Test)

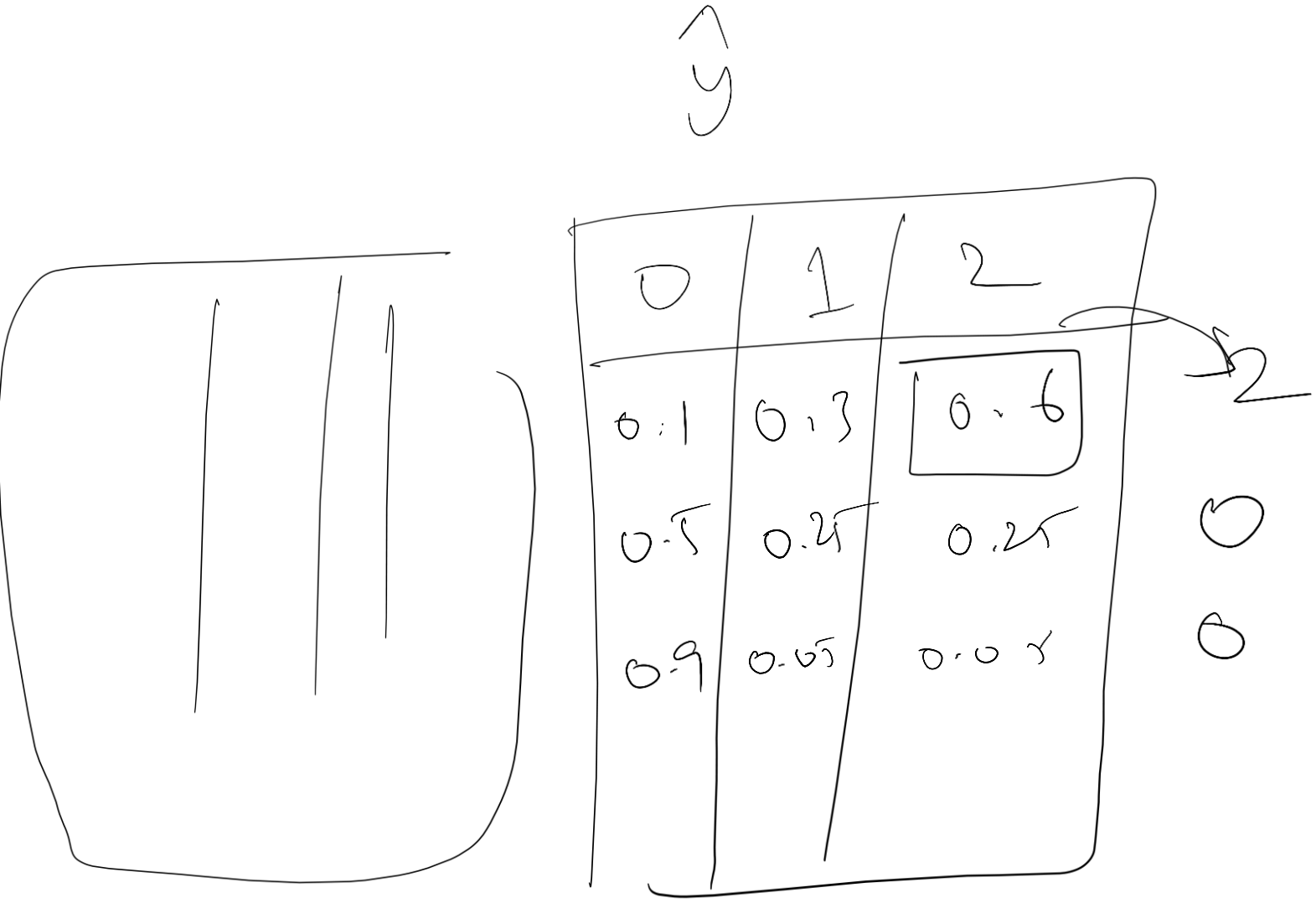
Model → RF, NB, XGB, Ada



0, 1, 2



Log Regression


ŷ
0
1
2
0.1
0.3
0.6
0.5
0.25
0.25
0.9
0.05
0.05
2
0
0

Br → 1 → Amazon

Brand	Search
Br 1	L 1
Br 2	M 2
Br 3	H 3
Br 4	

Brand

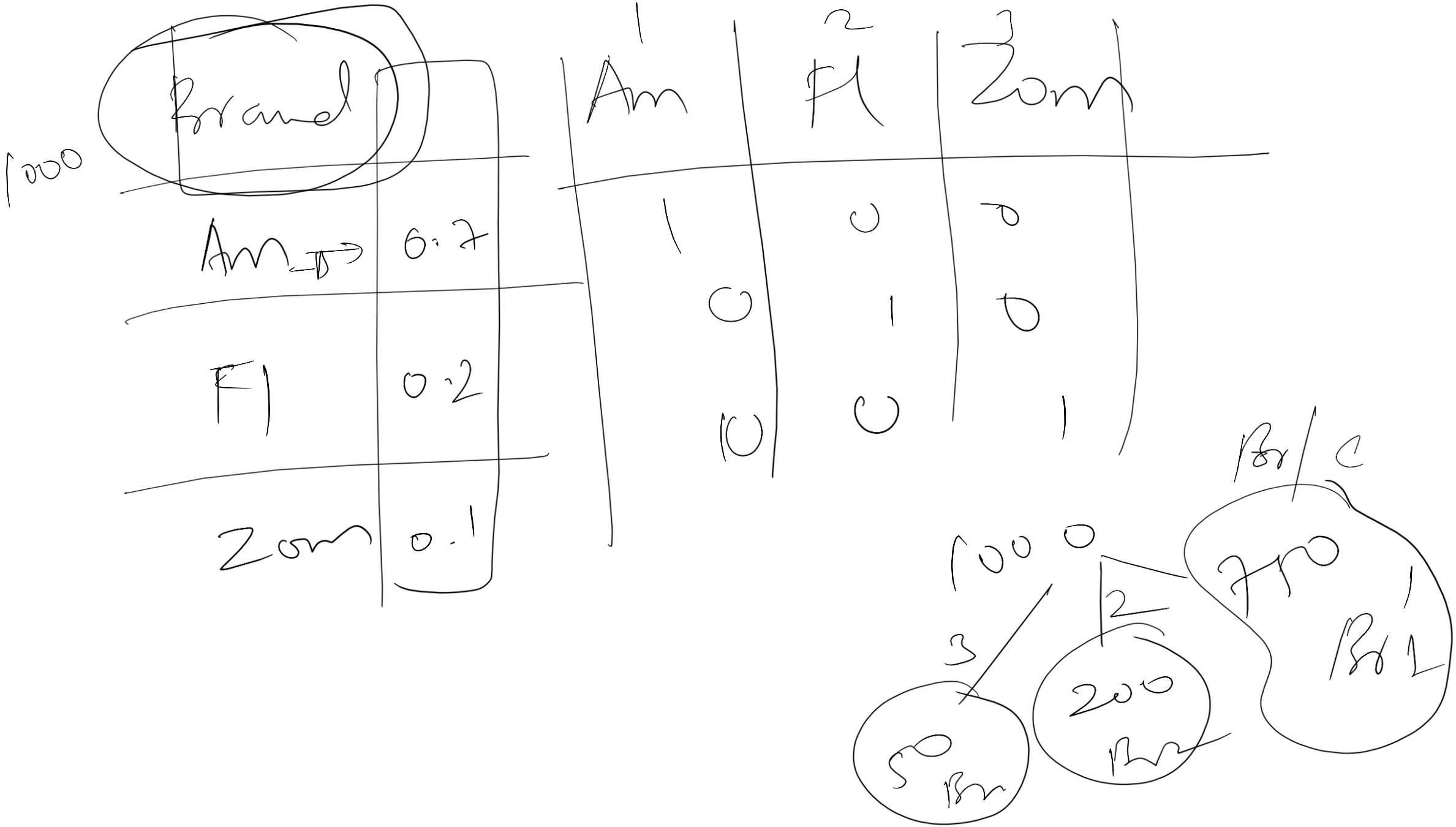
- Br 1 ⇒ 0.7
- Br 2 0.1
- Br 3 0.1
- Br 4 0.1

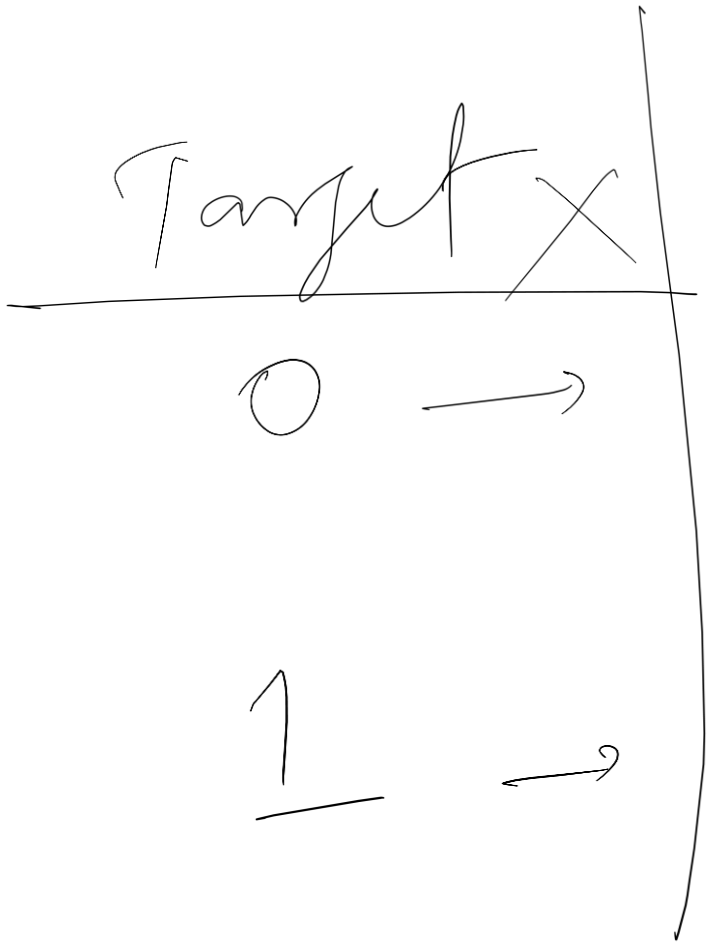
0.7 ✓

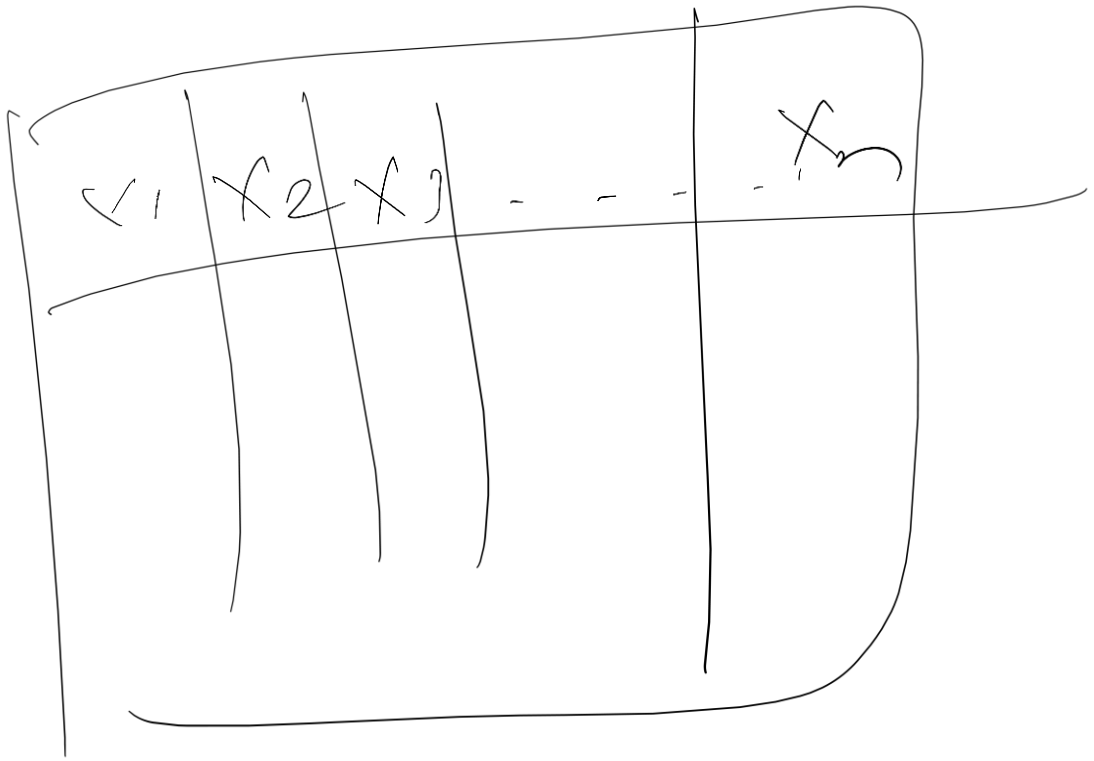
0.1

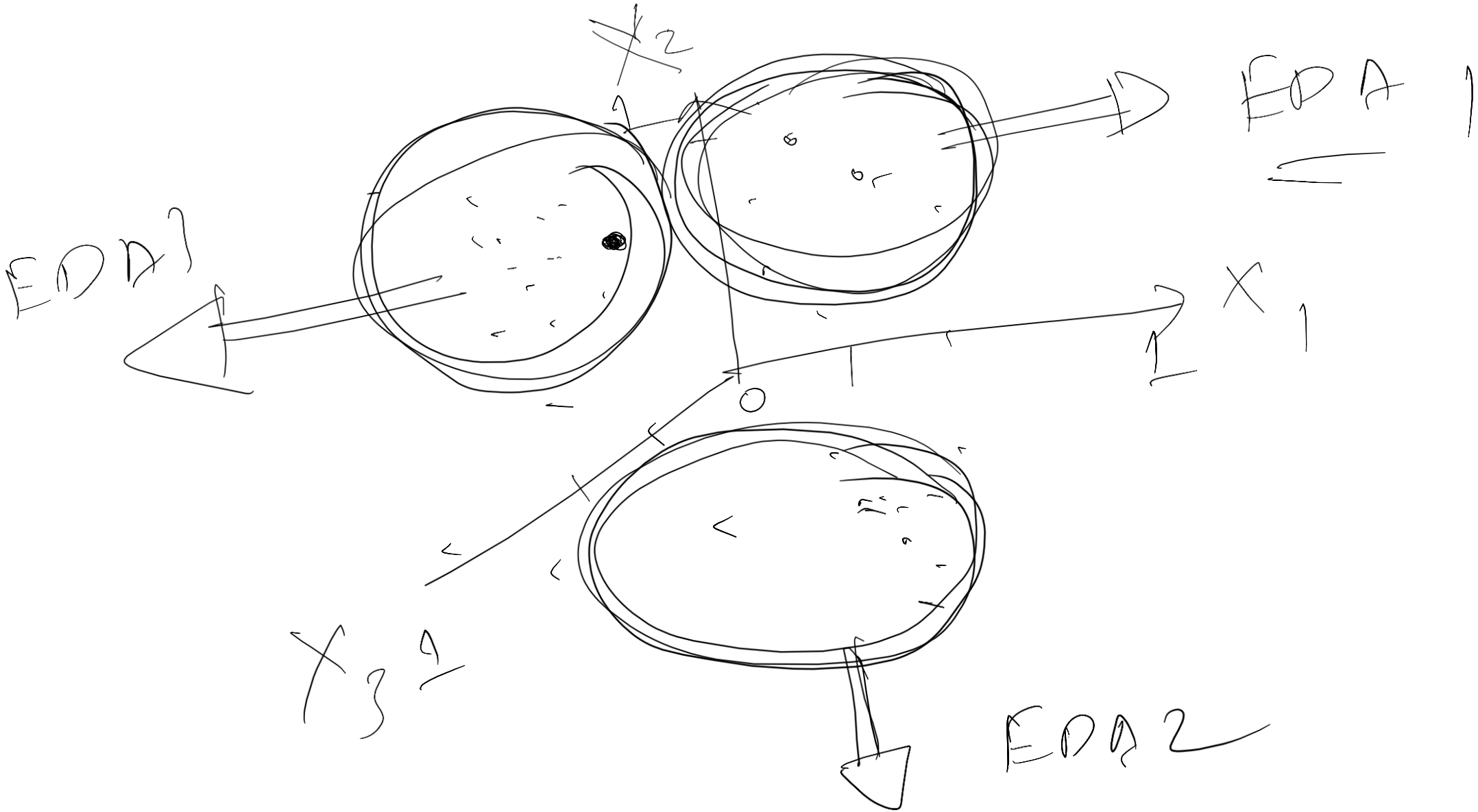
0.1

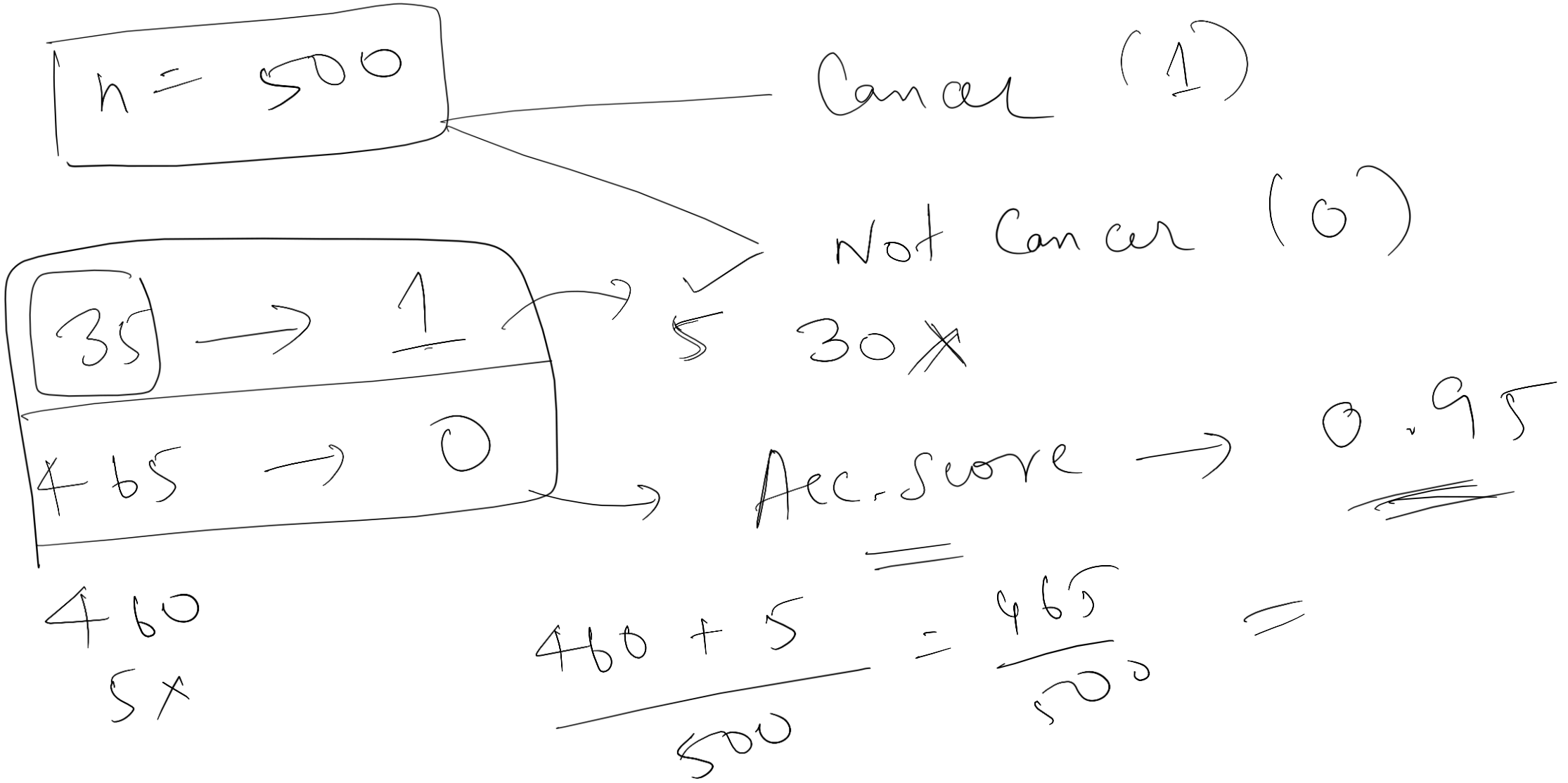
0.1


Brand
1000
Am → 0.7
F1 0.2
Zom 0.1
Am
F1
Zom
1 0 0
0 1 0
0 0 1
1
2
3
Br/C
Br1
700
1000
200
500

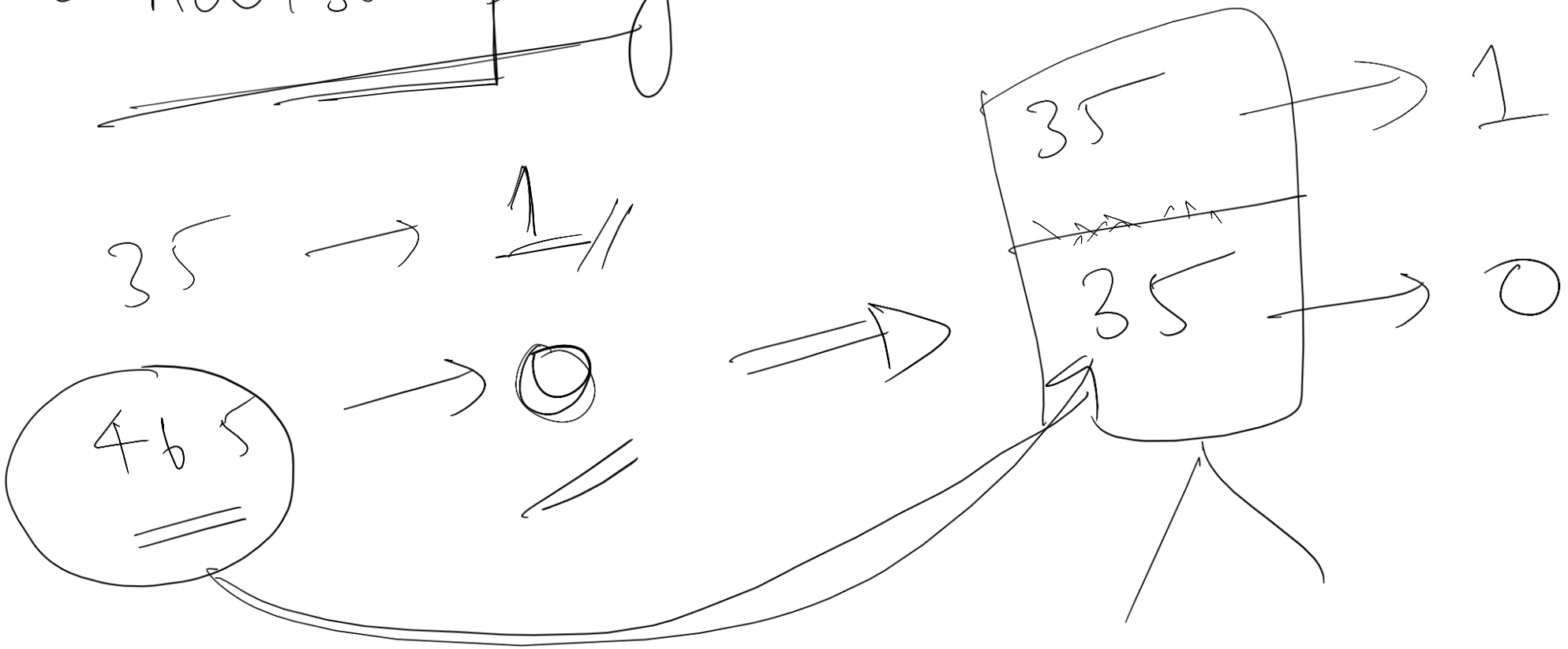

Target x
0 →
1 →


x1 x2 x3 - - - - xn


EDA1
X2
EDA 1
X1
O
X3
EDA2


n = 500
Cancer (1)
Not Cancer (0)
35 → 1
465 → 0
5
30 ✗
460
5 ✗
Acc. Score → 0.95
460 + 5 / 500 = 465 / 500 =

Undersampling:



465 → 1

465 → 0