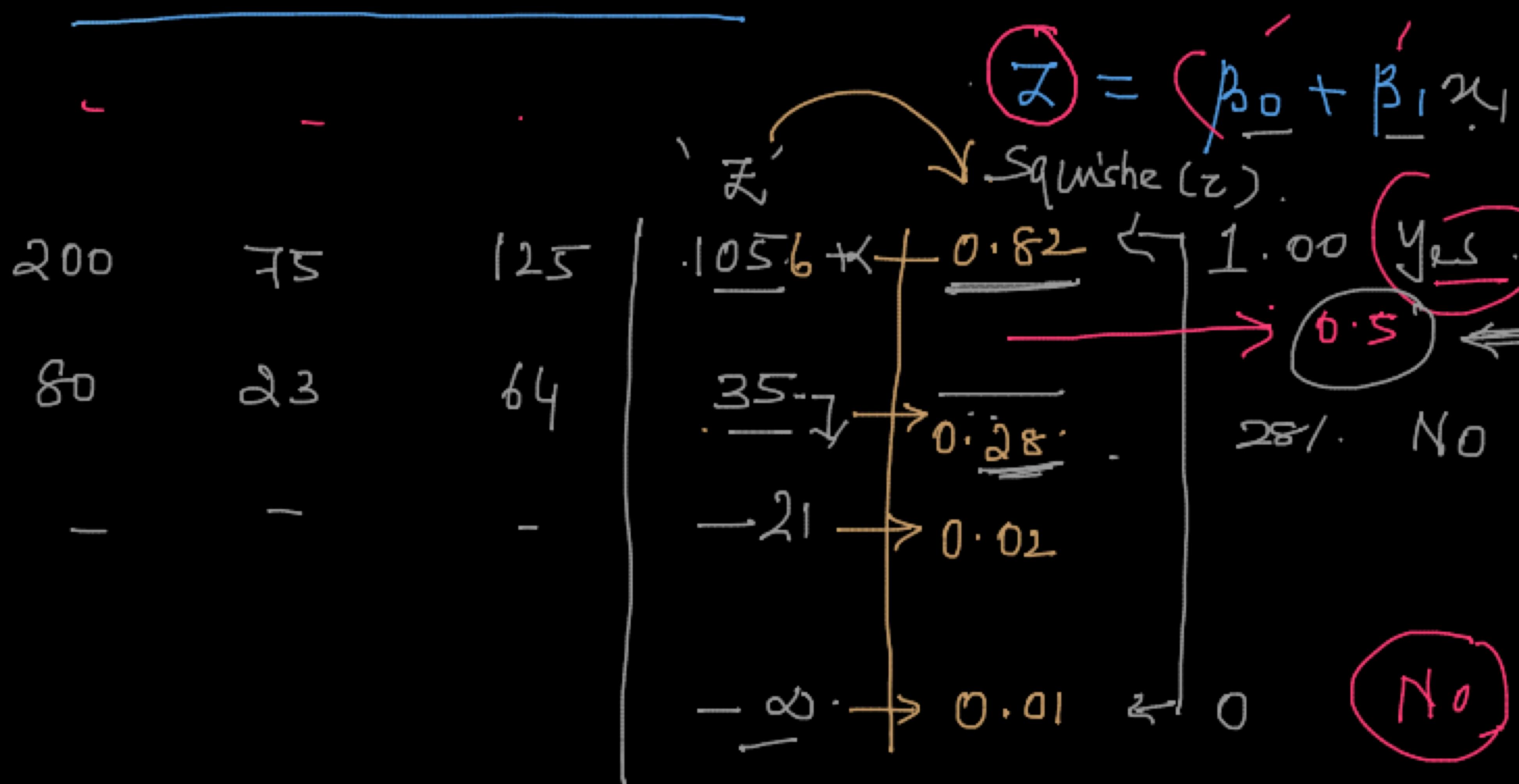


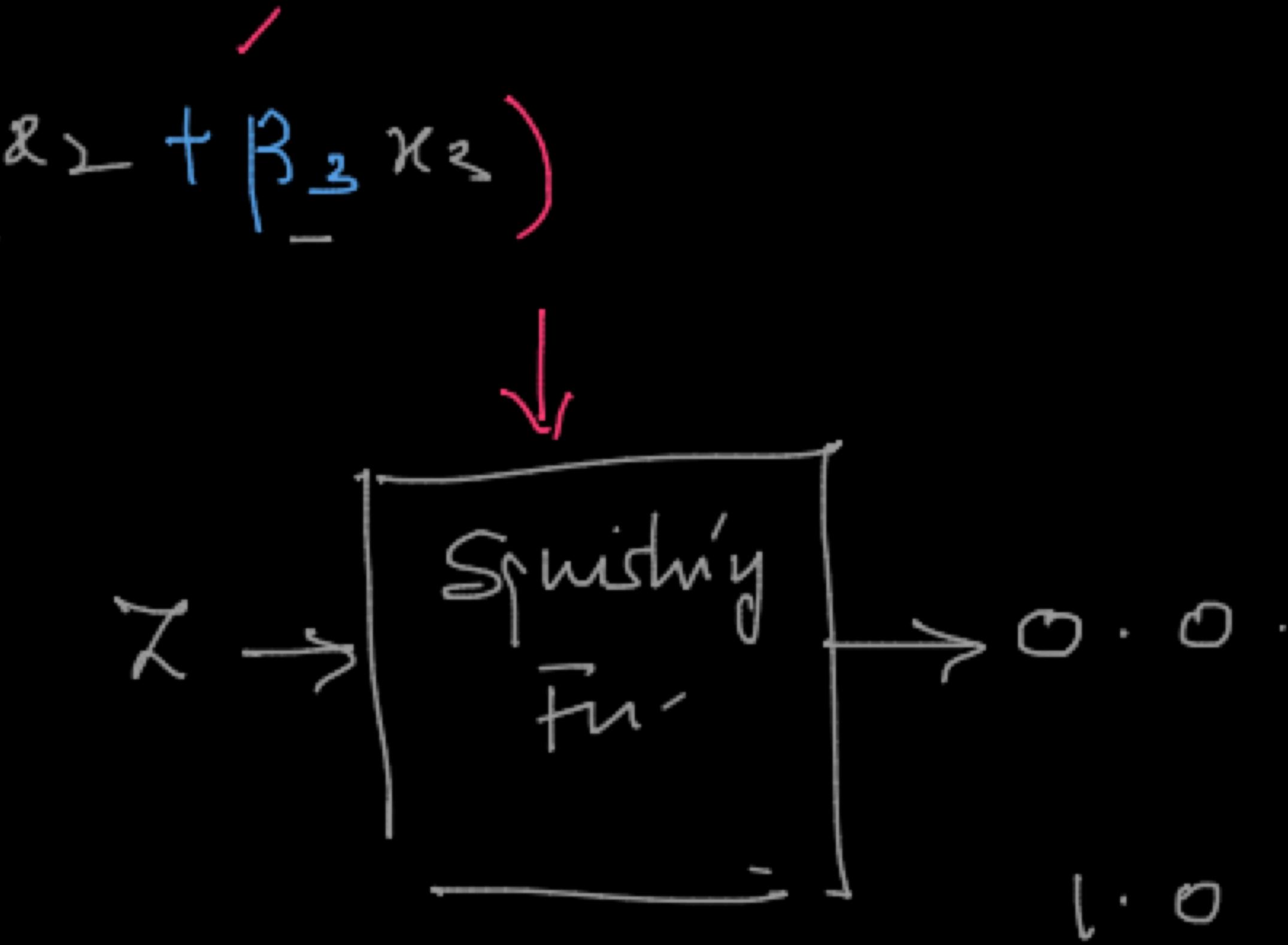
Logistic Regression

- classification Model
- y is categorical
- linear Model

x_1	x_2	x_3	
BP	cholesterol	WC	y
200	75	125	
80	23	64	
-	-	-	



Model



$$f_h(z) \Rightarrow 0.0 \quad 1.0$$

\hookrightarrow Sigmoid Fn.

x_1	x_2	x_3	\bar{x}	P_{+}	y
200	95	123	1200	0.8	Yes
80	23	64	-56	0.1	No
$Z = -\bar{x}$	+0	+0	\downarrow		
(Z)	$0 \cdot 0$	$0 \cdot 5$	$1 \cdot 0$	\checkmark	

$$z = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3$$

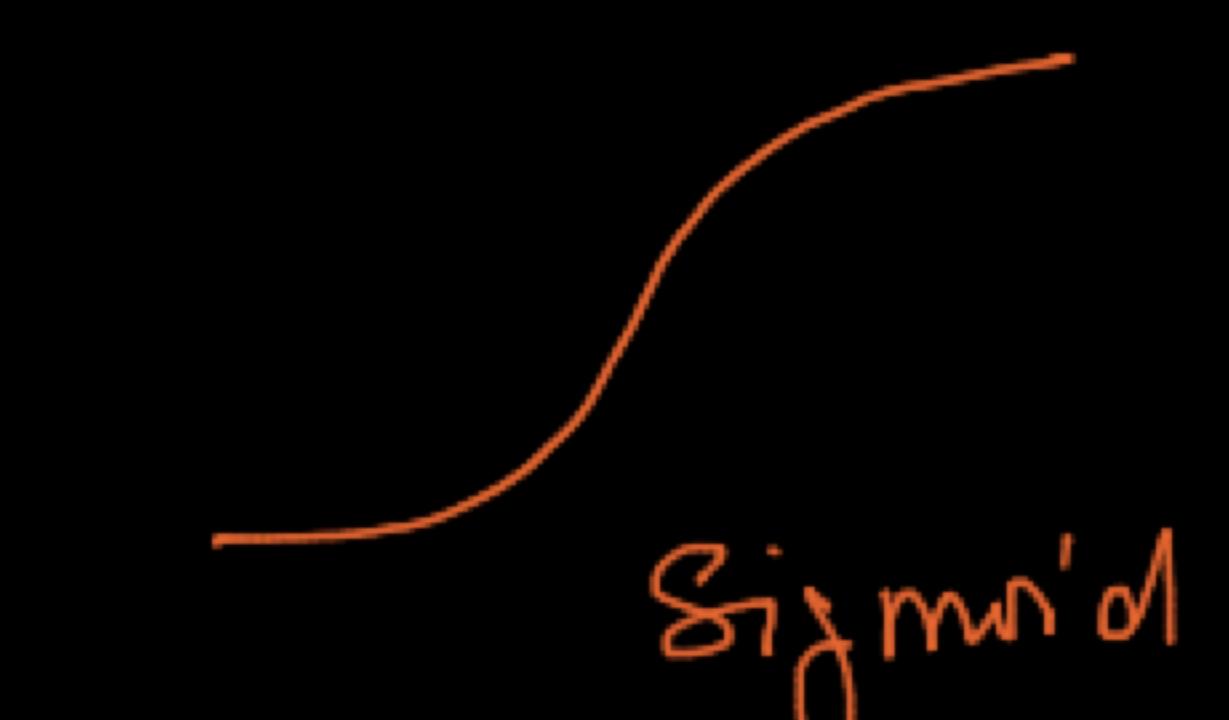
\rightarrow to \leftarrow

0.5
Threshold

ad. 902

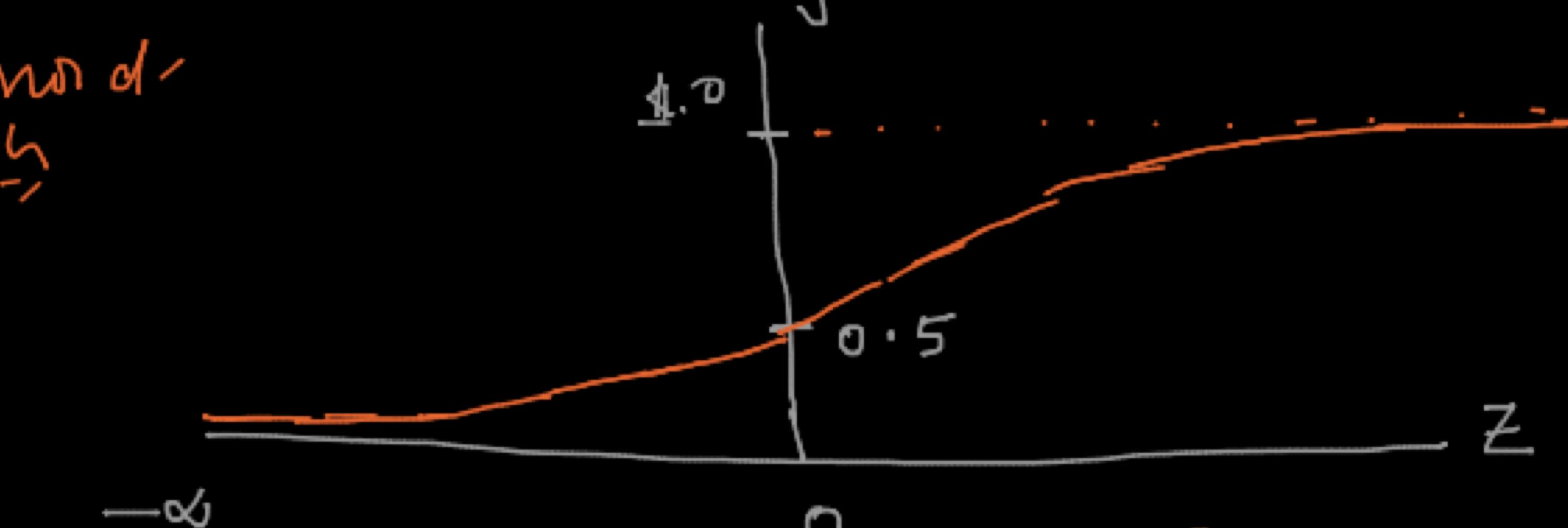
$$z \rightarrow \begin{bmatrix} f_n \\ \vdots \\ +\infty \end{bmatrix}$$

$$\rightarrow \{0, 0, \} \quad \text{Probability}$$



$$\text{Sigmoid}(z) = \frac{1}{1 + e^{-z}}$$

$$g(z) = \frac{1}{1 + \frac{1}{e^z}} = \frac{e^z}{e^z + 1}$$



$$x = +\infty$$

$$g(z) = \frac{e^z}{1+e^z} = 1.0$$

$$Z = -\infty \cdot \frac{1}{0} \Rightarrow \sqrt{\omega} \approx 0$$

$$g(z) = \frac{e}{1 + e^{-z}}$$

$$\frac{\partial}{\partial x} = 0$$

$$\mathcal{Z} = 0 \quad , \quad e^0 \quad .$$

$$g(2) = \frac{e}{1+e^0} = \frac{e}{1+1} = 0.5$$

	x_1	x_2	x_3	y	\hat{y}
1	-	-	-	0	0 ✓ TN
2	-	-	-	0	0 → FP
3	-	-	-	0	1 ✓ → FP
4	-	-	-	0	0 ✓ TN
5	-	-	-	1	0 → FN
6	-	-	-	1	1 ✓ → TP

$$\begin{aligned} TN &= 2 \checkmark & TP &= 1 \leftarrow \\ FP &= 0 \cdot & FN &= 1 \rightarrow \end{aligned}$$

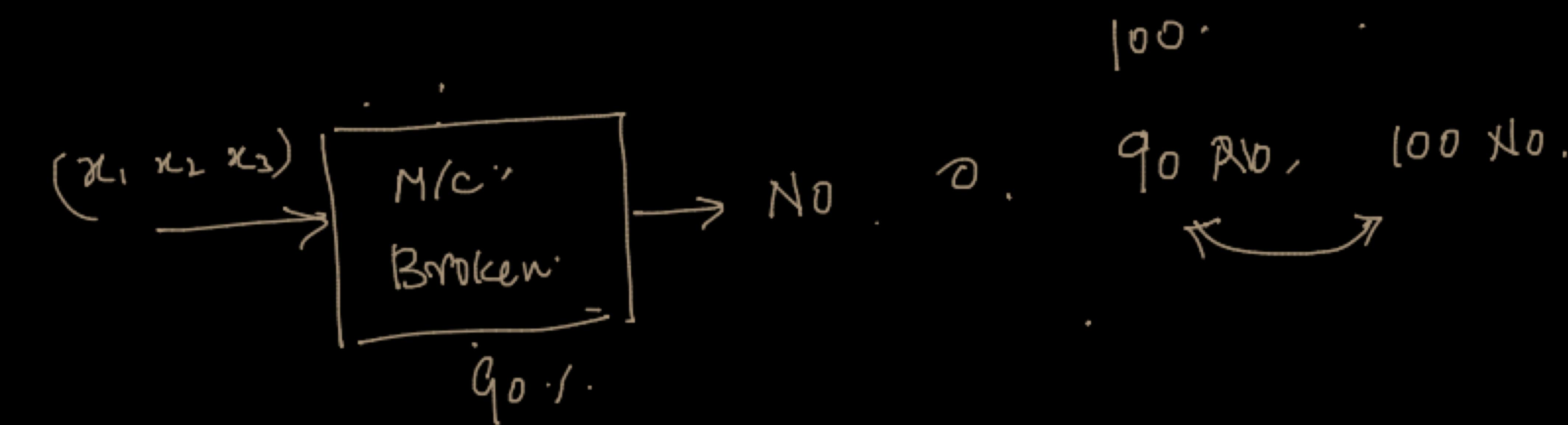
$$\begin{array}{r} 0 \quad 0 \\ 0 \quad 1 \\ 0 \quad 0 \\ 0 \quad 0 \\ 0 \quad 0 \\ \hline 0 \quad 0 \\ 0 \quad 0 \\ 0 \quad 0 \\ 1 \quad 0 \end{array} \quad \begin{array}{l} 50\% \\ \rightarrow 90\% \end{array}$$

$$\text{Accuracy} = \frac{4}{6} = \frac{50\%}{66.66\%}$$

Highly Imbalanced Dataset -

$$\begin{array}{l} 0 \approx 50\% \\ 1 \approx 50\% \end{array} \quad \text{Balanced} \quad \frac{4}{6}$$

$$\begin{array}{l} 0 \approx 90\% \\ 1 \approx 10\% \end{array} \quad \text{Imbalanced Dataset}$$



Confusion Matrix

		Ground Truth		
		True	False	
Predicted	True	TP ✓	FP	Precision = $\frac{TP}{TP+FP}$
	False	FN	TN	Recall = $\frac{TP}{TP+FN}$ (TPR, Sensitivity)

$$\text{Accuracy} = \frac{TP + TN}{TP + FP + FN + TN}$$

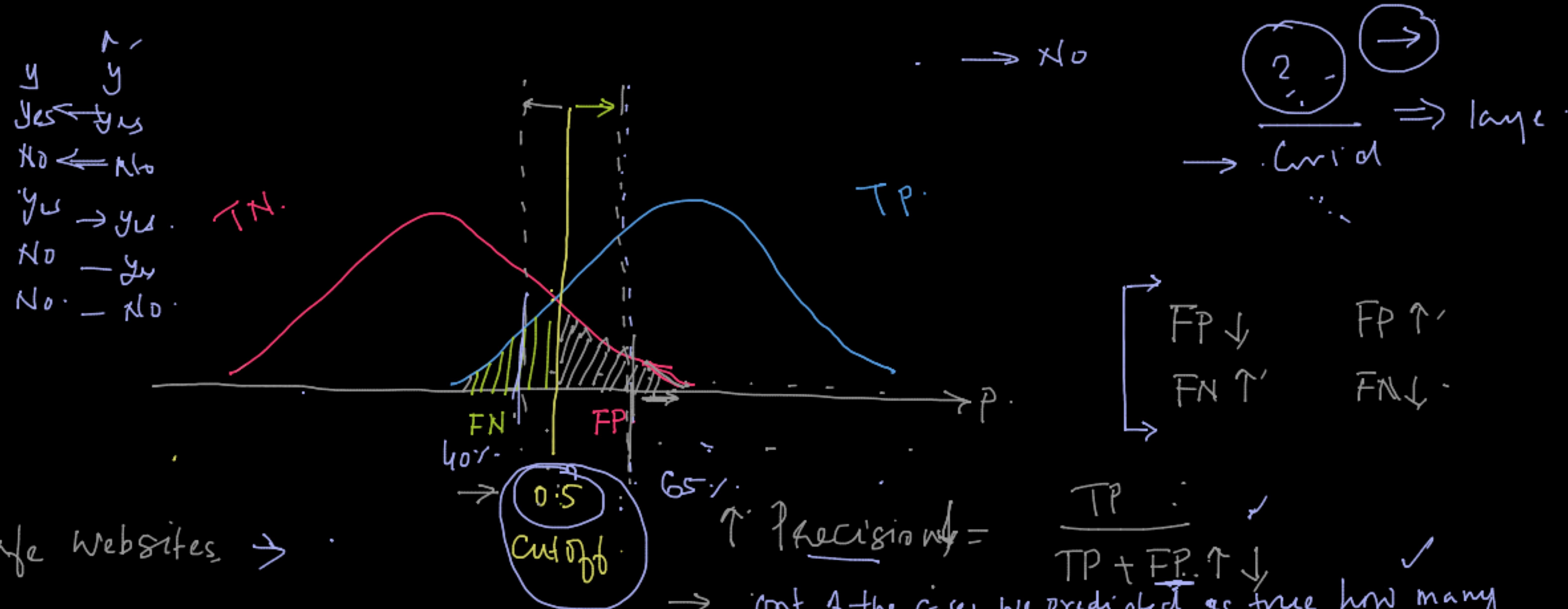
All predicted as true
Cases who actually have
the diseases.

Actual → | | | | | | | | | | → 3 7
No disease disease

We predict → | | | | | | | | | |

$$\text{Precision} = \frac{4}{5} \leftarrow$$

$$\text{Recall} = \frac{4}{7}$$



1. Safe Websites →

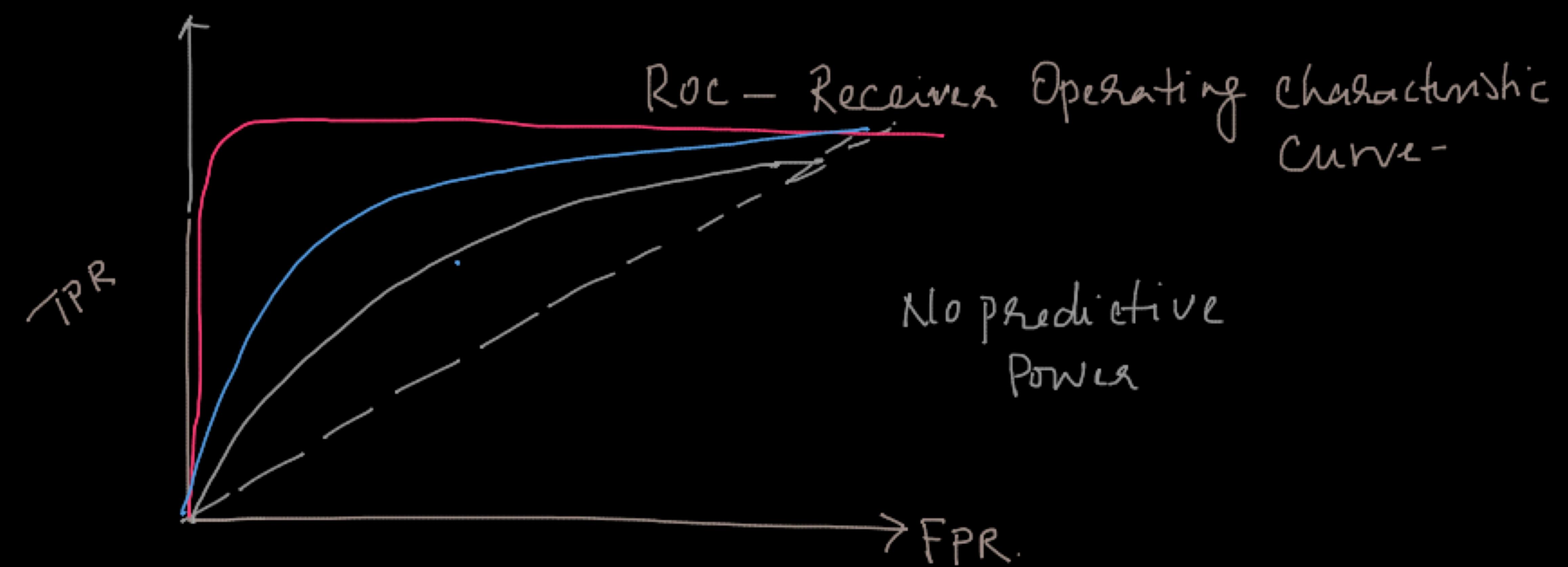
✓ Precision → Out of sites we predicted as safe how many were actually safe.

(Recall) → Out of sites that were actually safe how many did we correctly say safe

\uparrow Precision = $\frac{TP}{TP + FP} \uparrow$ ✓
 → Out of the cases we predicted as true how many were actually true.

\downarrow Recall = $\frac{TP}{TP + FN} \uparrow$ ✓

— Out of cases that were actually how many did we correctly predict.



AUC \rightarrow ROC

Regression

R^2

MSE / RMSE

MAE / MAPE

Classification

Accuracy / F1 score

Recall

Precision

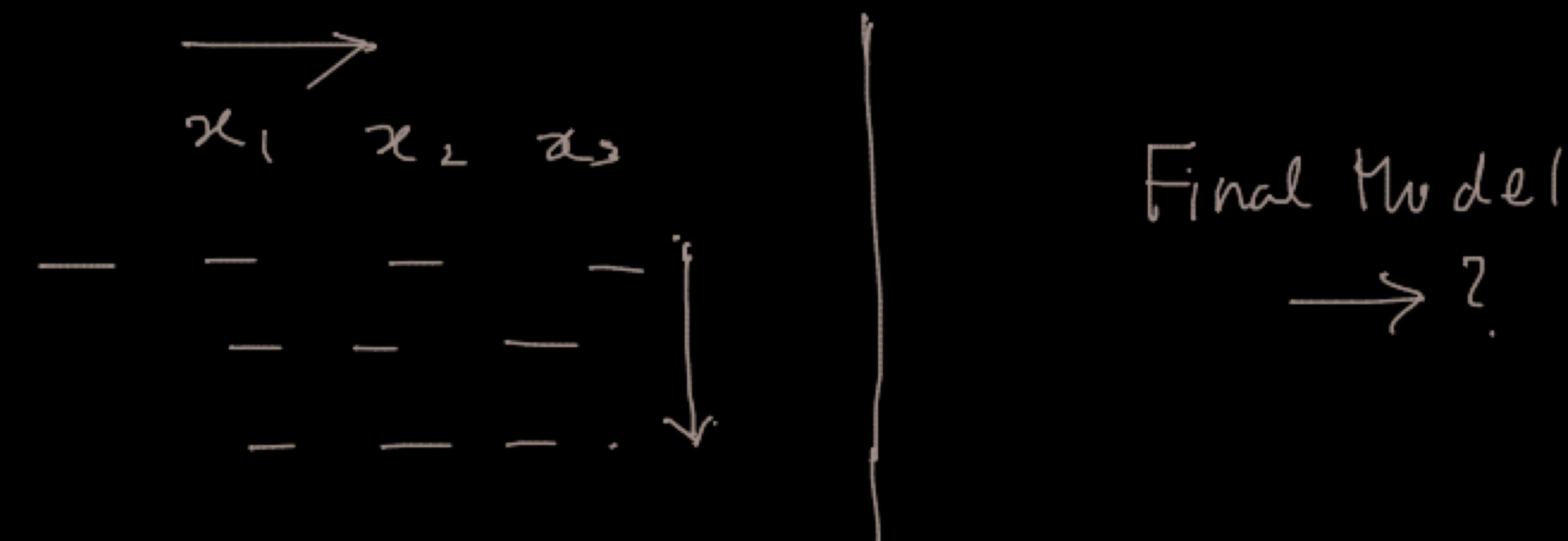
AUC - ROC

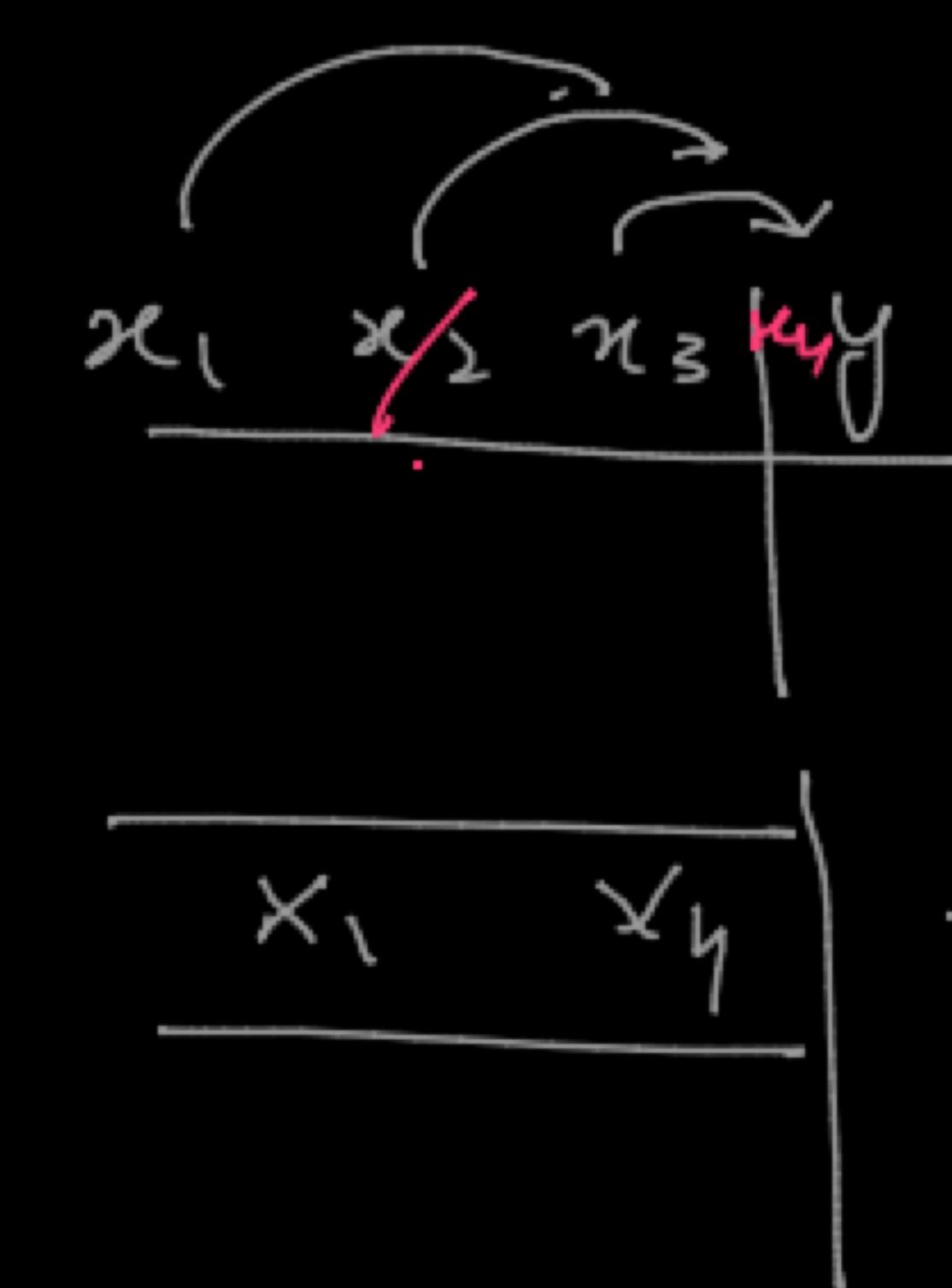
1. Predicting spam ✓ →

2. Risky cases →

Imbalanced Dataset → Both Acc / F1-score

$$\text{F1-Score} = \frac{2 (\text{Precision} \times \text{Recall})}{(\text{Precision} + \text{Recall})}$$





1. What model to build.

→ Regression / classification
y-continuous y - categorical.

Regression.

1. Can we build linear model?

a. Scatter Plot



Feature Engineering

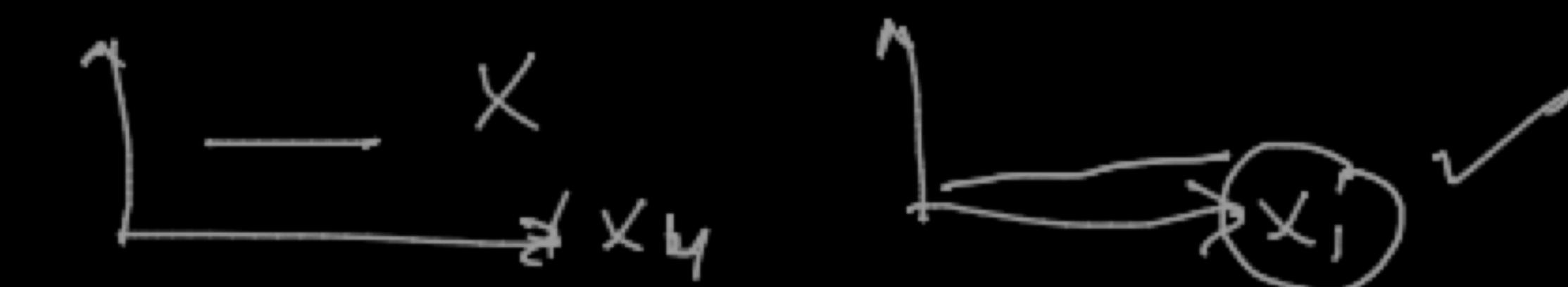
b. find the correlation coefficient

$$r \geq 0.85$$

2. Is there a relation between x-values.

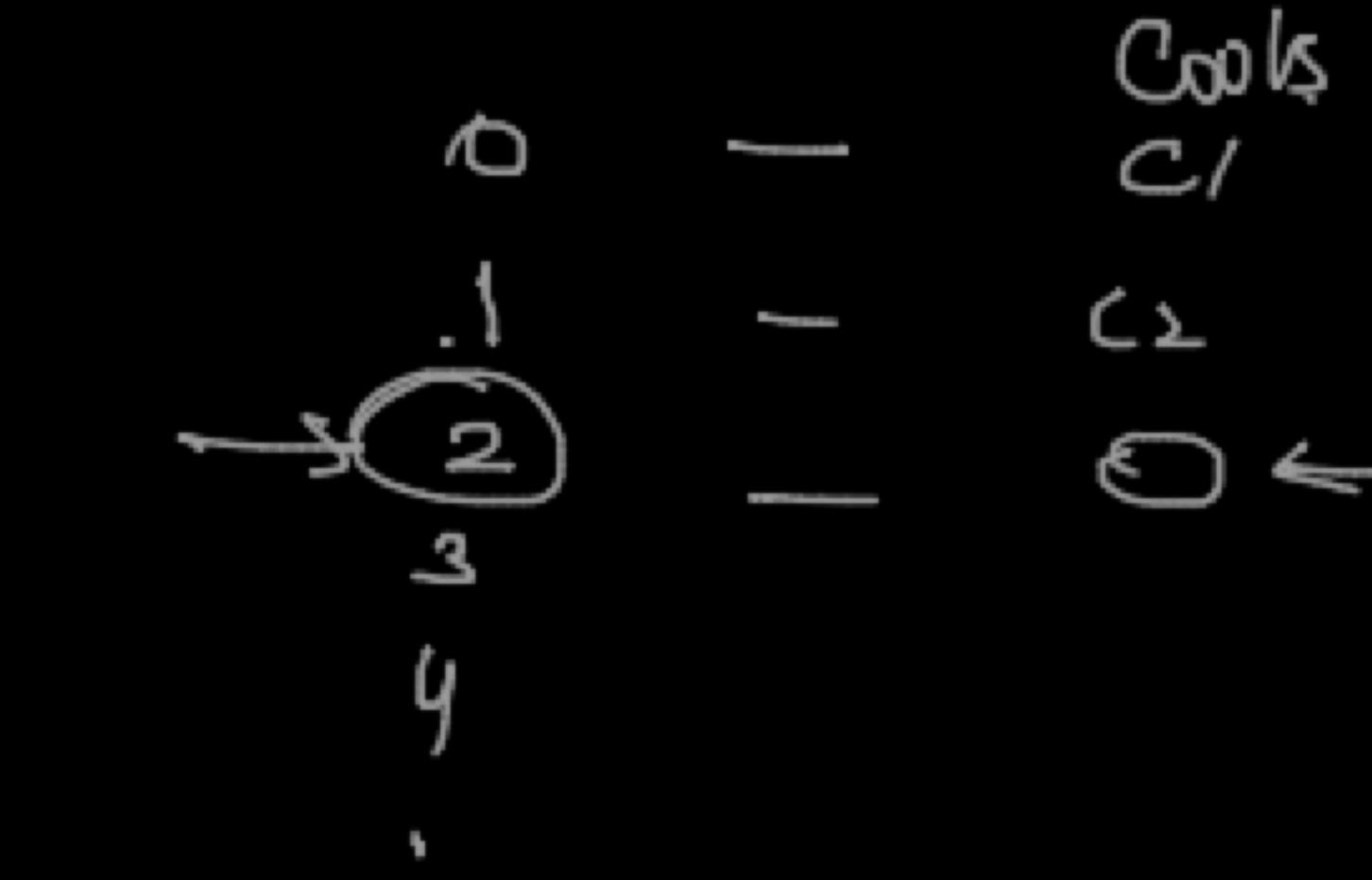
$$(a) \ x_1 \ x_4 = \frac{R^2}{y \leftarrow x_1} \quad y \leftarrow x_4$$

(c) Partial Regression



$$(b) VIF = x_1 \ x_3 \ x_4$$

$$\frac{x_1 \rightarrow x_3 \ x_4 - R^2}{x_3 \rightarrow} = \frac{1}{1-R^2} = V_{IF-x_1} \text{ Smallest } V_{IF} \rightarrow$$



Any Records / datapoints - influential?

- Cook's distance
 - Inflation
- Remove data points.

outliers, High leverage → Influential

x_1	x_4	y
-	-	.
-	-	.
-	-	.
-	-	.
-	-	.
-	-	.
-	-	.

train-test-split

- Low Training Error
High Testing Error
overfitting

R^2
 MSE , $RMSE$
 MAE

High Training Error
Testing Error high

x_1	x_4	x_5	y
1	1	1	1

→ deployment

→ Accuracy / F1
Precision
Recall
AUC - ROC