

Classification

Cat  $\gamma_6 \rightarrow 0.5$  — target Variable is categorical  
0.8

— Logistic Regression

— Linear Model

<u>BP</u>	<u>Chol</u>	<u>Sugar</u>	<u>H1</u>	<u>Z</u> → P	<u>y</u>
180	110	300	Y ✓	0.92	Y ✓
80	20	90	N ✗	0.50	N ✗
			N	0.21	N 10
			Y	0.75	Y ✗
			<u>Z</u>	$\Rightarrow N$	

$$y = f(x) \quad (ax_1 + bx_2 + cx_3)$$

↳ linear Eqn

$$\hat{y} = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3$$

Revr-n

$$\Rightarrow Z = \beta_1 BP + \beta_2 \times Chol + \beta_3 \times Sugar \quad \leftarrow$$

$\infty$   
 $\ell \rightarrow \infty$

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

$+ \infty$

$$e^{-\infty} \rightarrow \frac{1}{\infty} \approx 0$$

$$= \frac{1}{1 + \frac{1}{e^z}}$$

$z \rightarrow -\infty$

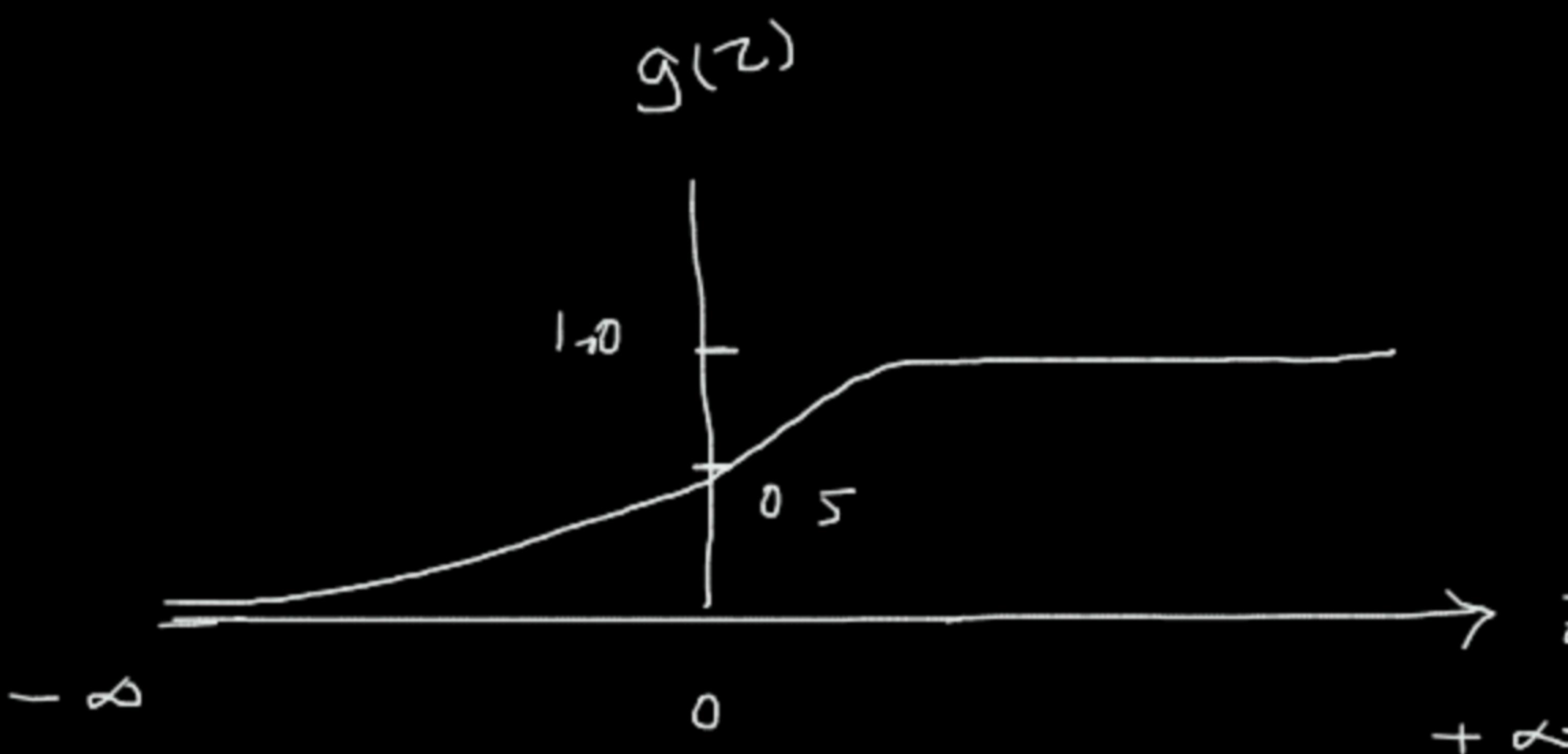
$$z \rightarrow \begin{cases} g(z) \\ 0 \end{cases} \rightarrow \begin{cases} y \\ 0.5 \end{cases}$$

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

$z \rightarrow +\infty$

$$g(z) = \frac{e^{-\infty}}{1 + e^{-\infty}} \Rightarrow 0 \quad z \rightarrow 0 \quad \frac{e^0}{1 + e^0} \rightarrow \frac{1}{1+1} = 0.5$$

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



Sigmoid

$g(z) \rightarrow$  Sigmoid Fn

	$x_1$	$x_2$	$x_3$	$y$	$\hat{y}$
1				y	y
2				N	N
3				N	y
4				y	N
5				y	y

$$\left( \frac{3}{5} \right)$$

Accuracy

↳ The percentage of correctly predicted values

Imbalanced Data

$$\begin{aligned} 100 &\rightarrow 80 \text{ N} - 50 \\ &\rightarrow 20 \boxed{y} - 50 \rightarrow \\ &\qquad\qquad M - 50 \end{aligned}$$

$$80 \rightarrow T_R \rightarrow \boxed{\begin{array}{cc} 64 & N \\ 16 & y \end{array}} \text{ Balanced} \quad \left\{ \begin{array}{ll} 45 & \text{+ve} \\ 48 & \text{-ve} \\ 7 & \text{Ne} \end{array} \right. \rightarrow$$

$$20 \rightarrow T_U \quad \boxed{\begin{array}{c} 15 \text{ N} \\ 5 \text{ y} \end{array}}$$

Resampling Techniques →  
Training Data

$$100 \rightarrow \boxed{MC} \rightarrow \begin{array}{l} Y - 10 \rightarrow N \\ N - q_0 - \end{array} \left( \frac{q_0}{100} \right)$$

F1-score pr.

Confusion Matrix

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

		Ground Truth	
		True (P)	False (N)
P	True (P)	TP ✓	FP
	False (N)	FN	TN ✓

Actual	Predicted
$y$	$\hat{y}$
$y$	$y \rightarrow \text{TP}$
$N$	$y \rightarrow \text{FP}$
$N$	$N \rightarrow \text{TN}$
$y$	$y \rightarrow \text{TP}$
$y$	$N \rightarrow \text{FN}$
$N$	$N \rightarrow \text{TN}$
$y$	$y \rightarrow \text{TP}$
$N$	$N \rightarrow \text{TN}$
$y$	$y \rightarrow \text{TP}$
$N$	$N \rightarrow \text{TN}$

Ground Truth			
$P$	$T_{\text{True}}(P)$	$Fals_{\text{e}}(N)$	
$N$	$T_{\text{True}}(P)$	$FP(1)$	
$N$	$Fals_{\text{e}}(N)$	$FN(1)$	$TN(4)$
$T$			

No } cases that are actually +ve (5)

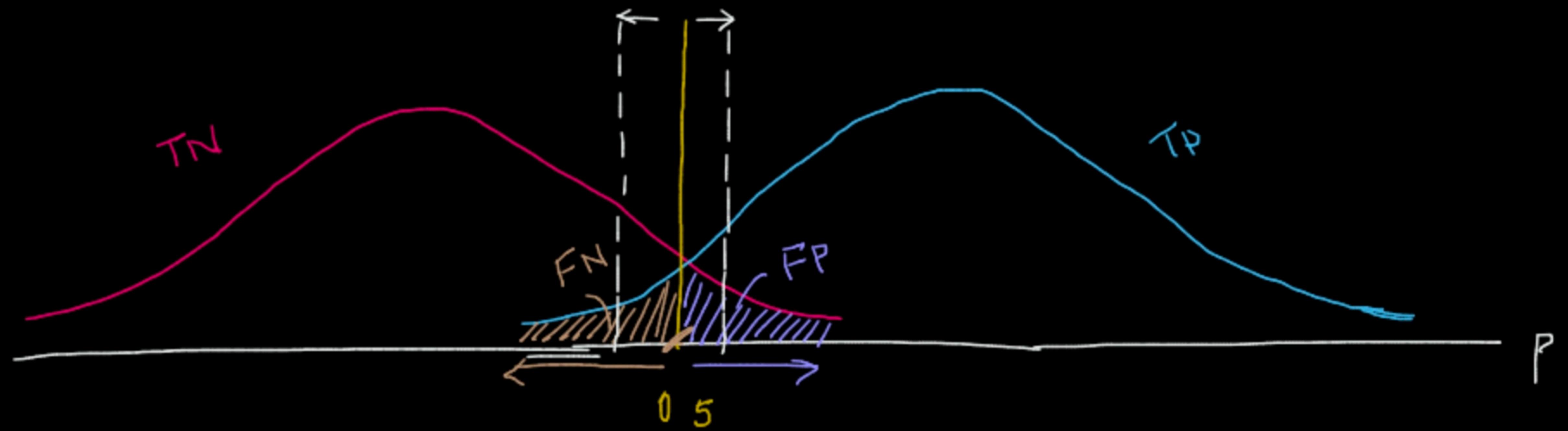
→ No of cases predicted as +ve (5)

$$\text{Accuracy} = \frac{8}{10} \rightarrow \frac{TP+TN}{TP+TN+FP+FN} \rightarrow \text{Percentage of correctly predicted cases}$$

Precision -  $\frac{TP}{TP+FP} = (4/5) \rightarrow$  out of the cases predicted as +ve how many are actually +ve ✓

Recall -  $\frac{TP}{TP+FN} = (4/5) \rightarrow$  out of the cases that are actually +ve how many were predicted as +ve

$$F1\text{-Score} = \frac{2 \cdot Pr \times Re}{Pr + Re}$$



$FN \uparrow$     $FP \downarrow$

$FN \downarrow$     $FP \uparrow$

$$\uparrow PV - \frac{TP}{TP + \boxed{FP}} \downarrow \rightarrow \frac{\overline{TP}}{\overline{TP} + \boxed{\overline{FN}}} \uparrow \rightarrow$$

1 Doctor treating covid cases  
Predict whether patient has covid or not

High Recall

Churn predicted

2 Security software which predicts whether a pg is safe or not

High Precision

Low  $\rightarrow (80 \rightarrow 78)$

