

BINARY CLASSIFICATION

STAT 432
DALPIAZ

MODEL FIT TO ESTIMATION

(x_i, y_i)

FROM

VALIDATION → EVAL

OR

TEST → REPORT

MODEL FIT TO TRAIN

y_i

ACTUAL

0
0
0
0
0
0
1
1
1
1

$C(x_i)$

PREDICTED

0
0
0
0
1
1
0
1
1
1

→ "POSITIVE"

→ "NEGATIVE"

CONFUSION MATRIX

		<u>ACT</u>	
		1	0
<u>PRED</u>	1		
	0		

POSITIONS OF ACT/PRED AND 0/1 COULD CHANGE !!!

<u>ACTUAL</u>	<u>PREDICTED</u>	
0	0	→ TN
0	0	
0	0	
0	0	
0	1	→ FP
0	1	
1	0	→ FN
1	1	
1	1	
1	1	→ TP

		<u>ACT</u>	
		1	0
<u>PRED</u>	1	3	2
	0	1	4

TP (True Positive) points to the cell (1,1) containing 3.
 FP (False Positive) points to the cell (1,0) containing 2.
 FN (False Negative) points to the cell (0,1) containing 1.
 TN (True Negative) points to the cell (0,0) containing 4.

P = 4 (Minority Class)
 N = 6 (Majority Class)

MINORITY CLASS MAJORITY CLASS

		<u>ACT</u>	
		1	0
<u>PRED</u>	1	3	2
	0	1	4

TP (True Positive) points to the cell (1,1) containing 3.
 FP (False Positive) points to the cell (1,0) containing 2.
 FN (False Negative) points to the cell (0,1) containing 1.
 TN (True Negative) points to the cell (0,0) containing 4.
 P = 4 (Total Positive Predictions) and N = 6 (Total Negative Predictions) are indicated below the table.

* $\text{Acc} = \frac{\text{TP} + \text{TN}}{\text{P} + \text{N}} = \frac{3 + 4}{4 + 6} = 0.7$ (1 - MISCLASS)

AND MANY MORE... (F_1 , MCC, ...)

SEE BOOK/WIKIPEDIA

$$P_{\text{REV}} = \frac{P}{P + N} = \frac{4}{4 + 6} = 0.4$$

$$S_{\text{ENS}} = \frac{\text{TP}}{P} = \frac{3}{4} = 0.75$$

$$S_{\text{PEC}} = \frac{\text{TN}}{N} = \frac{4}{6} = 0.66\bar{6}$$

$$P_{\text{PV}} = \frac{\text{TP}}{\text{TP} + \text{FP}} = \frac{3}{3 + 2} = 0.6$$

$$F_{\text{DR}} = \frac{\text{FP}}{\text{FP} + \text{TP}} = \frac{2}{2 + 3} = 0.4$$

No

INFORMATION RATE



PROPORTION OF
MAJORITY CLASS

$$NIR = \max \left\{ \frac{P}{P+N}, \frac{N}{P+N} \right\}$$

PREV

1-PREV

IF $ACC > NIR$ → REASONABLE CLASSIFIER

IF $ACC < NIR$ → USELESS CLASSIFIER

$$0.7 > 0.6 \quad \checkmark$$

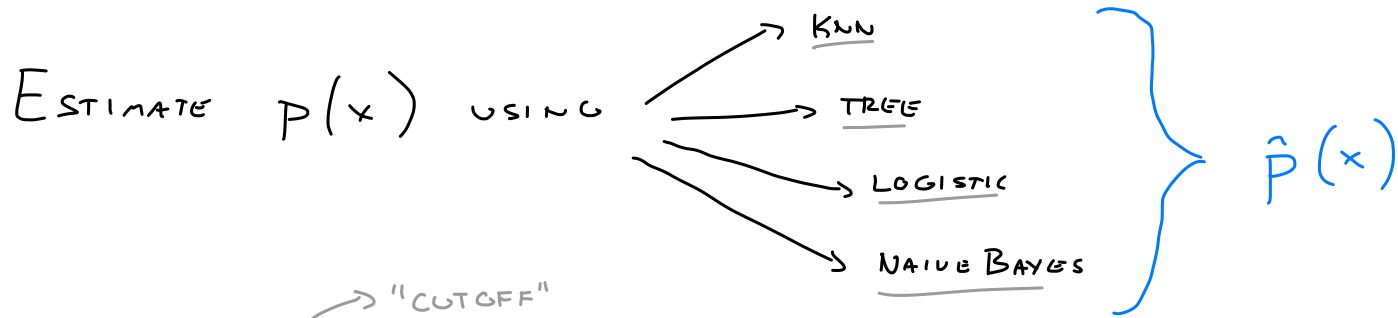
$$\underline{Y=0} \quad \text{or} \quad \underline{Y=1}$$

$$p(x) \triangleq P[Y=1 | X=x]$$

$$1 - p(x) = P[Y=0 | X=x]$$

$$\rightarrow P[Y=1 | X=x] \geq P[Y=0 | X=x]$$

$$C^B(x) = \begin{cases} 1 & p(x) \geq 0.5 \\ 0 & p(x) < 0.5 \end{cases}$$



SET $0 \leq \alpha \leq 1$ "CUTOFF"

$$C_{\alpha}(x) = \begin{cases} 1 & \hat{P}(x) \geq \alpha \\ 0 & \hat{P}(x) < \alpha \end{cases}$$


ASSUMED $\alpha=0.5$

AS $\alpha \uparrow$ HARDER TO CLASSIFY AS $Y=1$

$$\text{CLASSIFIER} = f(\hat{P}(x), \alpha)$$

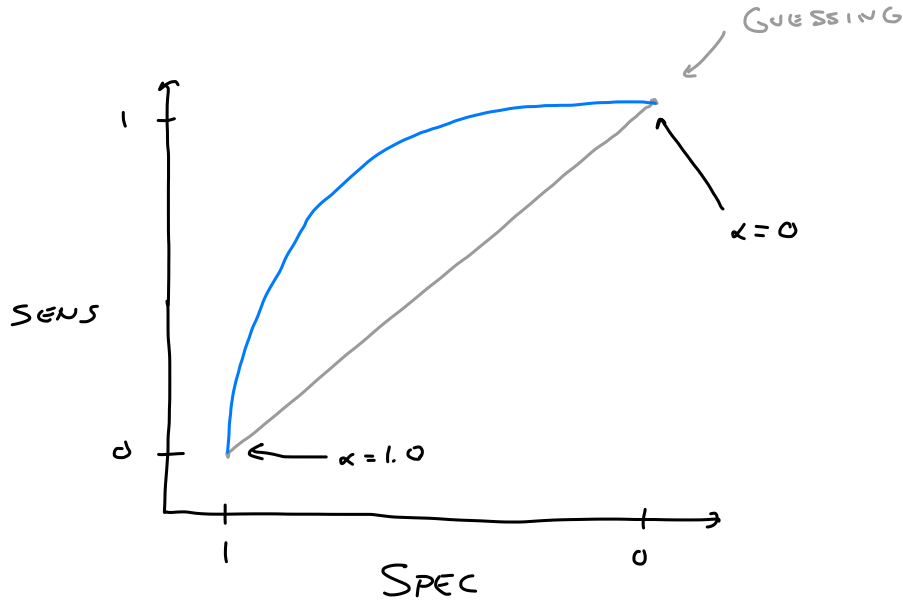
y_i	$\hat{p}(x_i)$	$C_{0.0}(x_i)$	$C_{0.25}(x_i)$	$C_{0.5}(x_i)$	$C_{0.75}(x_i)$	$C_{1.0}(x_i)$
0	0.1	1	0	0	0	0
0	0.1	1	0	0	0	0
0	0.2	1	0	0	0	0
0	0.3	1	1	0	0	0
0	0.6	1	1	1	0	0
0	0.7	1	1	1	0	0
1	0.4	1	1	0	0	0
1	0.7	1	1	1	0	0
1	0.8	1	1	1	1	0
1	0.9	1	1	1	1	0

$TP/P = SENS =$	1.00	1.00	0.75	0.50	0.00
$TN/N = SPEC =$	0.00	0.50	0.66	1.00	1.00
$ACC =$	0.40	0.70	0.70	0.80	0.60


 EXPECTED TO BE BEST

EVALUATE $\hat{p}(x)$ INSTEAD OF $C(x)$? ROC CURVE!

INPUTS
 y
 $\hat{p}(x)$



AUC

- Bigger = Better
- 1 = PERFECT
- 0.50 = "WORST"