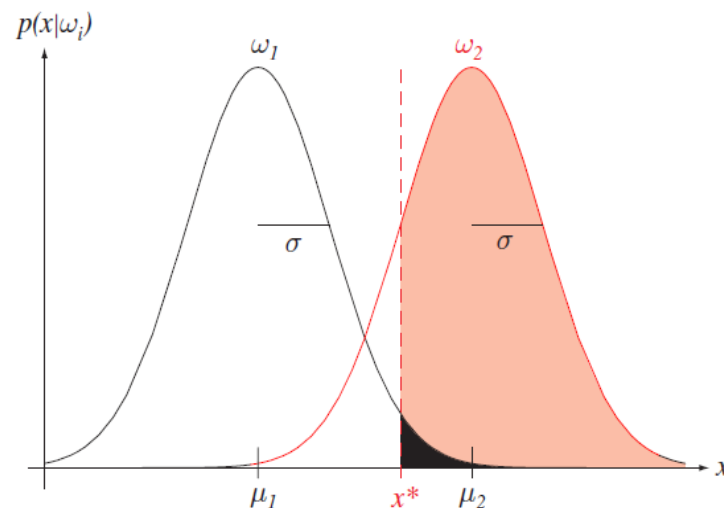


# Classification measures (binary case)

# 2 types of truth and errors

- Predicting the possession of a disease by genetic diagnosis.
- In the figure below, if we use separating hyperplane marked by  $x^*$ , then we get the two types of truth;
  - Predicting a sick person as sick.
  - Predicting a healthy person as healthy.
- In the figure below, if we use separating hyperplane marked by  $x^*$ , then we get the two types of truth;
  - Predicting a sick person as healthy.
  - Predicting a healthy person as sick.



# example

The following table is often called as contingency table.

		actual	
		Yes	No
prediction	Yes	80 ( <b>T</b> True <b>P</b> Positive)	20 ( <b>F</b> False <b>P</b> Positive)
	No	10 ( <b>F</b> False <b>N</b> Negative)	90 ( <b>T</b> True <b>N</b> Negative)

# Evaluating classifier performance based on Contingency Table

		actual	
		Yes(Class1)	No(Class2)
prediction	Yes(Class1)	80 ( <b>T</b> True <b>P</b> Positive)	20 ( <b>F</b> False <b>P</b> Positive)
	No(Class2)	10 ( <b>F</b> False <b>N</b> Negative)	90 ( <b>T</b> True <b>N</b> Negative)

**Accuracy: ACC** =  $(TP+TN)/(TP+FP+TN+FN) = (TP+TN) / \text{All} = 170/200$

**False Positive Rate : FPR** =  $FP/(FP+TN) = 20/110$

**True Positive Rate(a.k.a. sensitivity, recall): TPR** =  $TP/(TP+FN)=80/90$

**Positive Predictive Value(a.k.a. precision): PPV** =  $TP/(TP+FP)=80/100$

# Other measures

"Cheat sheet" on accuracy, precision, recall, TPR, FPR, specificity, sensitivity, ROC, and all that stuff!

William H. Press, ver 1.0, 3/29/08

Confusion matrix:

		actual	
		+	-
classifier	+	TP	FP <small>Type I error</small>
	-	FN <small>Type II error</small>	TN
	column totals:	P	N

		actual	
		+	-
classifier	+	TP	FP
	-	FN	TN
		accuracy (ACC)	

		actual	
		+	-
classifier	+	TP	FP
	-	FN	TN
		neg. predictive value (NPV)	

		actual	
		+	-
classifier	+	TP	FP
	-	FN	TN
		specificity (SPC)	

↕ "one minus" ↕

		actual	
		+	-
classifier	+	TP	FP
	-	FN	TN
		false pos. rate (FPR)	

ROC curve: FPR (x) vs. TPR (y)

precision-recall curve: TPR (x) vs. PPV (y)

		actual	
		+	-
classifier	+	TP	FP
	-	FN	TN
		pos. predictive value (PPV) ≡ precision	

↕ "one minus" ↕

		actual	
		+	-
classifier	+	TP	FP
	-	FN	TN
		false discovery rate (FDR)	

value (between 0 and 1) = numerator / denominator

numerator = dark color shade

denominator = dark + light color shade

blue: value 1 is good

pink: value 0 is good

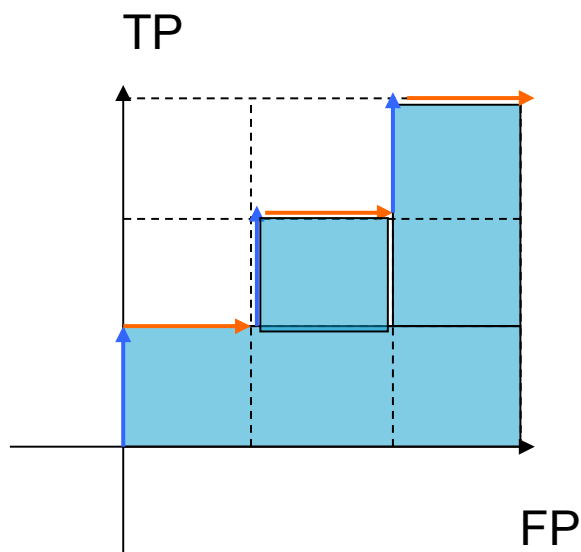
Map points from ROC to  
Precision-Recall or vice-versa:  
(TPR same values in both)

$$PPV = \frac{P \cdot TPR}{P \cdot TPR + N \cdot FPR} \quad (ROC \text{ to } P-R)$$

$$FPR = \frac{P \cdot (1 - PPV) \cdot TPR}{N \cdot PPV} \quad (P-R \text{ to } ROC)$$

# Receiver Operator Characteristic (ROC) and Area Under the Curve (AUC)

- ROC curve takes FPR as x-axis, and TPR as y-axis.
- AUC is an area under the ROC curve



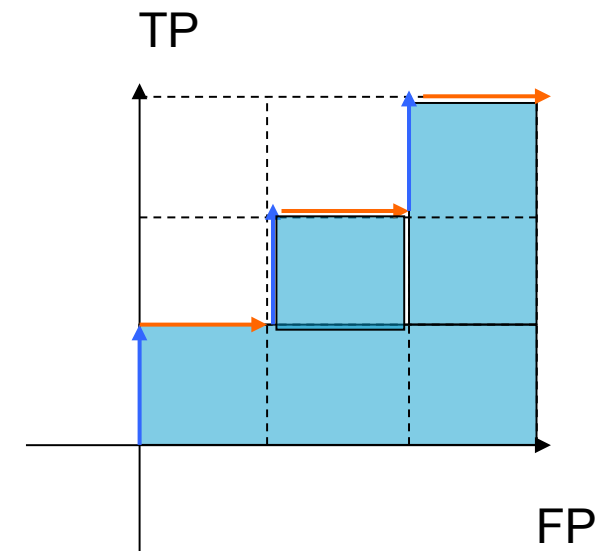
$$AUC = 6/9 = 0.66$$

$$(0 \leq AUC \leq 1)$$

# Computing AUC

- Sort the prediction results in a descending order.
- By looking at score from the top one by one, then move rightward if FP, and upward if TP.
- Calculate the relative area under the ROC curve.

Predicted score	True Class	FP	TP
0.9	+	0	1
0.7	-	1	1
0.3	+	1	2
-0.1	-	2	2
-0.4	+	2	3
-0.6	-	3	3

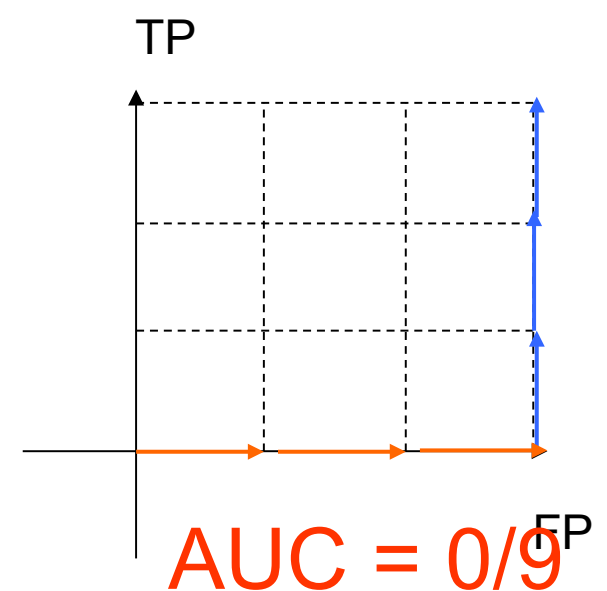
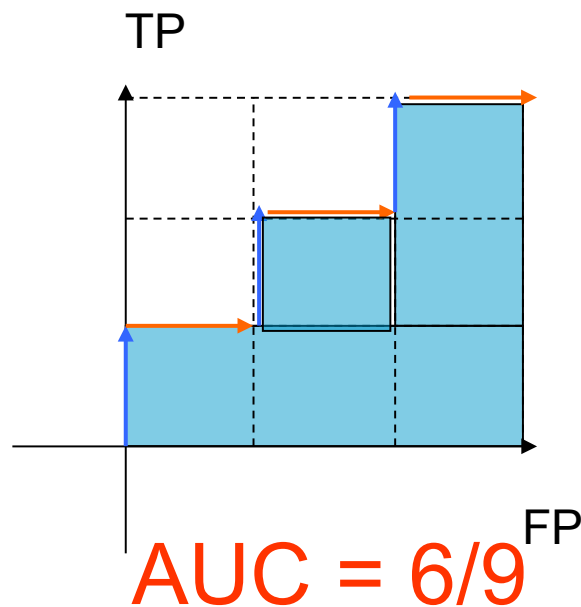
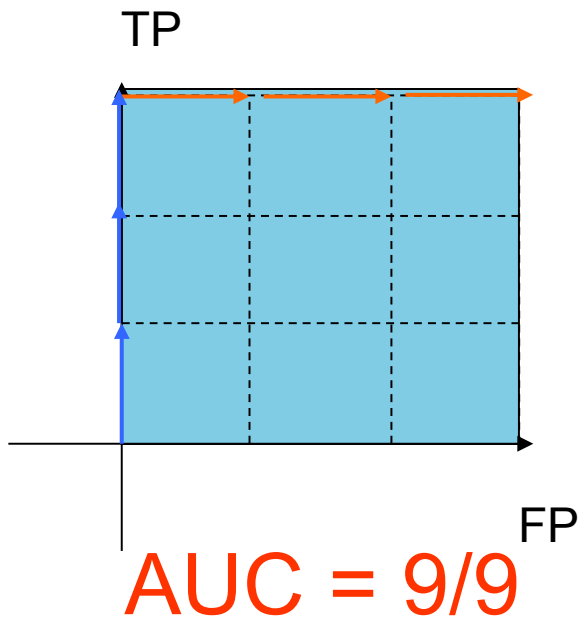


$$AUC = 6/9 = 0.66$$

$$(0 \leq AUC \leq 1)$$

# AUC(Area Under the Curve)

- $0 \leq \text{AUC} \leq 1$
- AUC curve located above the other AUC curves suggests better performance.





## Ex. 2

- Compute Accuracy, Sensitivity, Specificity, Recall, Precision, and FDR from the following table.

		actual	
		Yes(Class1)	No(Class2)
prediction	Yes(Class1)	300	15
	No(Class2)	50	1000

# Ex. 3

- Calculate AUC score from the following table.

Prediction	True Class
1.0	+
0.8	+
0.5	+
0.3	-
0.1	+
-0.1	-
-0.2	+
-0.6	-
-0.9	-
-1.0	-

(Ex. 4) previous exercise revisited.

- Complete the following table by evaluating all the digits drawn by 200 people in test set T.
- Which digit is often misclassified ? To which digit ?

[illegible]