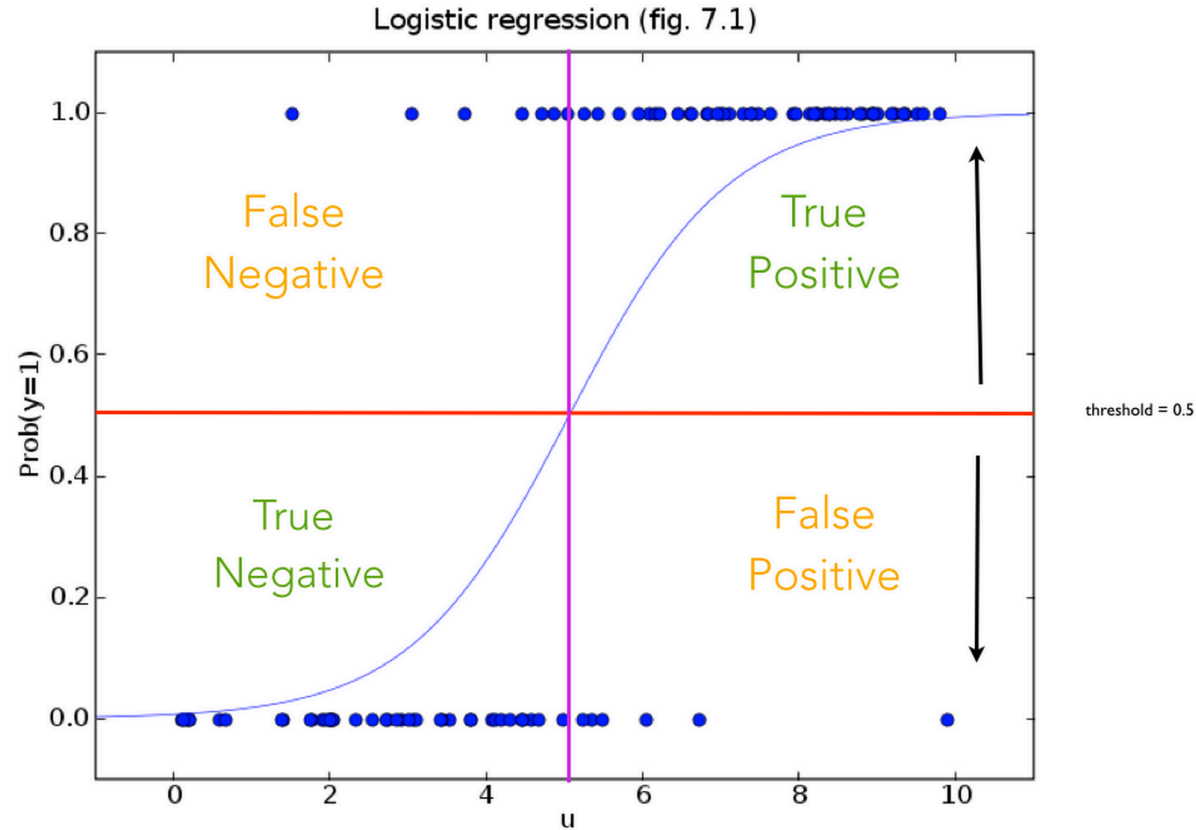


Profit Curves

Overview

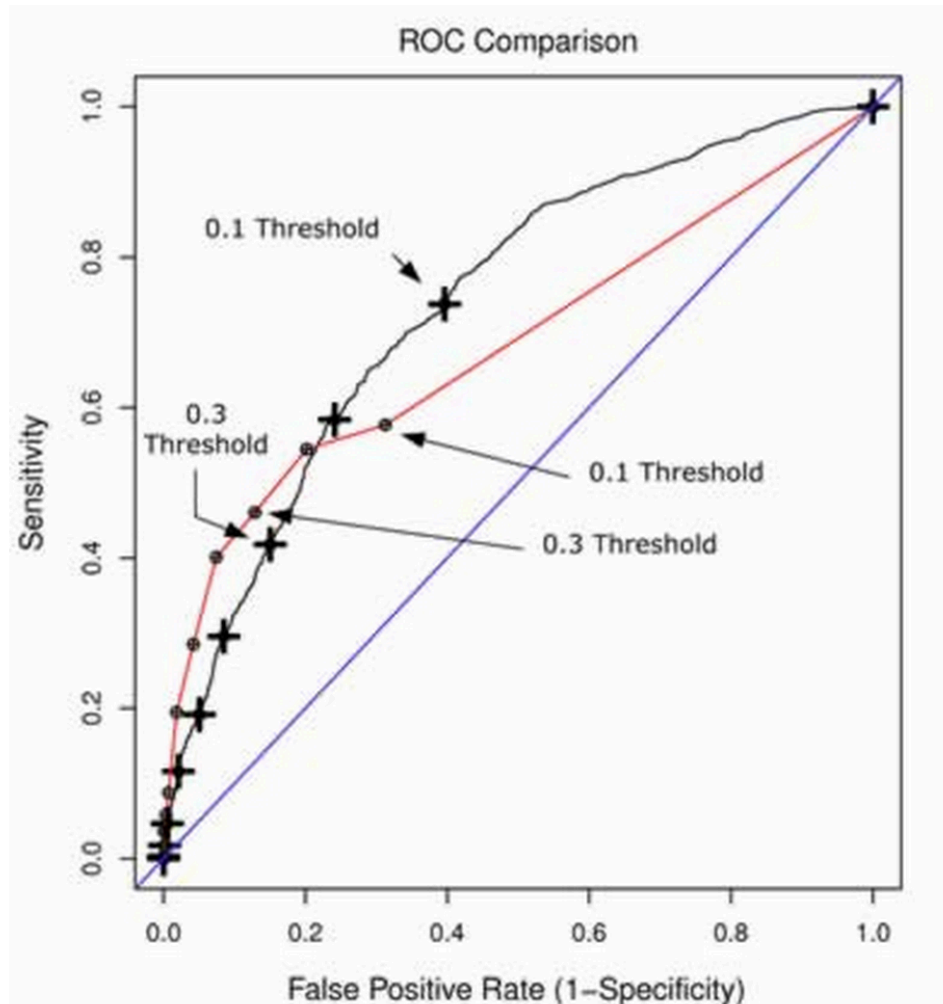
- Review of Logistic Regression and ROC Curves
- Profit Calculation
 - Simple
 - Expanded
- Profit Curves

Logistic Regression



	Predicted Yes	Predicted No
Actual Yes	True positive	False negative
Actual No	False positive	True negative

Logistic Regression - Evaluation



- Area under the Curve (AUC)
- F1 Score
- Precision / Recall
- Sensitivity / Specificity

Evaluation of Classifier

true positive (TP)

eqv. with hit

true negative (TN)

eqv. with correct rejection

false positive (FP)

eqv. with false alarm, Type I error

false negative (FN)

eqv. with miss, Type II error

accuracy (ACC)

$$ACC = (TP + TN) / (P + N)$$

F1 score

is the harmonic mean of precision and sensitivity

$$F1 = 2TP / (2TP + FP + FN)$$

sensitivity or true positive rate (TPR)

eqv. with hit rate, recall

$$TPR = TP / P = TP / (TP + FN)$$

specificity (SPC) or true negative rate (TNR)

$$SPC = TN / N = TN / (FP + TN)$$

precision or **positive predictive value (PPV)**

$$PPV = TP / (TP + FP)$$

negative predictive value (NPV)

$$NPV = TN / (TN + FN)$$

fall-out or false positive rate (FPR)

$$FPR = FP / N = FP / (FP + TN)$$

false discovery rate (FDR)

$$FDR = FP / (FP + TP) = 1 - PPV$$

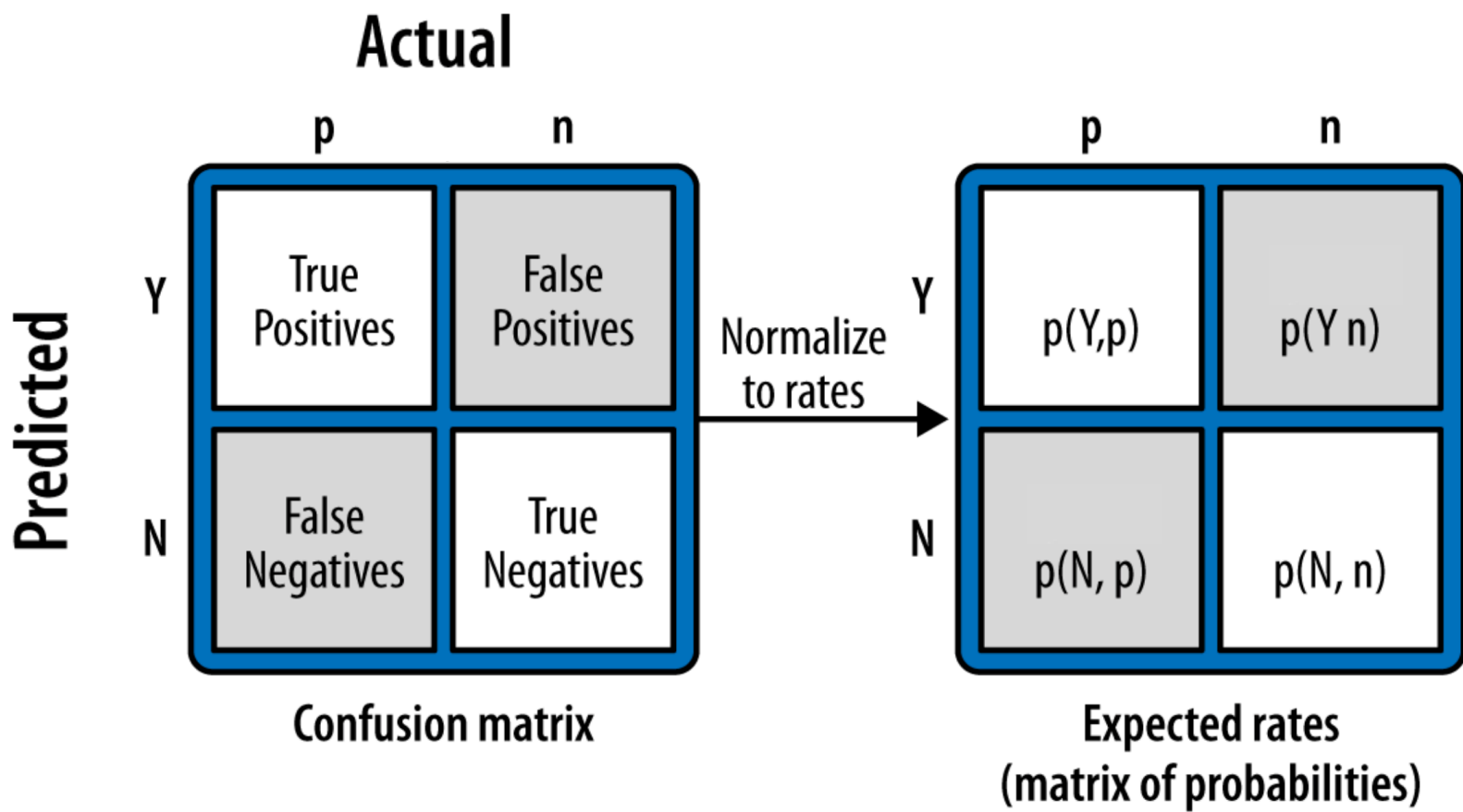
false negative rate (FNR)

$$FNR = FN / (FN + TP) = 1 - TPR$$

Confusion Matrix

- ROC Curves assume equal cost to misclassification
 - However different errors have different costs
 - Correct classification can have different benefits

		Actual	
		p	n
Predicted	Y	True Positives	False Positives
	N	False Negatives	True Negatives



Cost benefit Matrix

- Defined from the business situation:

		Actual	
		p	n
Predicted	Y	$b(Y,p)$	$c(Y,n)$
	N	$c(N,p)$	$b(N,n)$

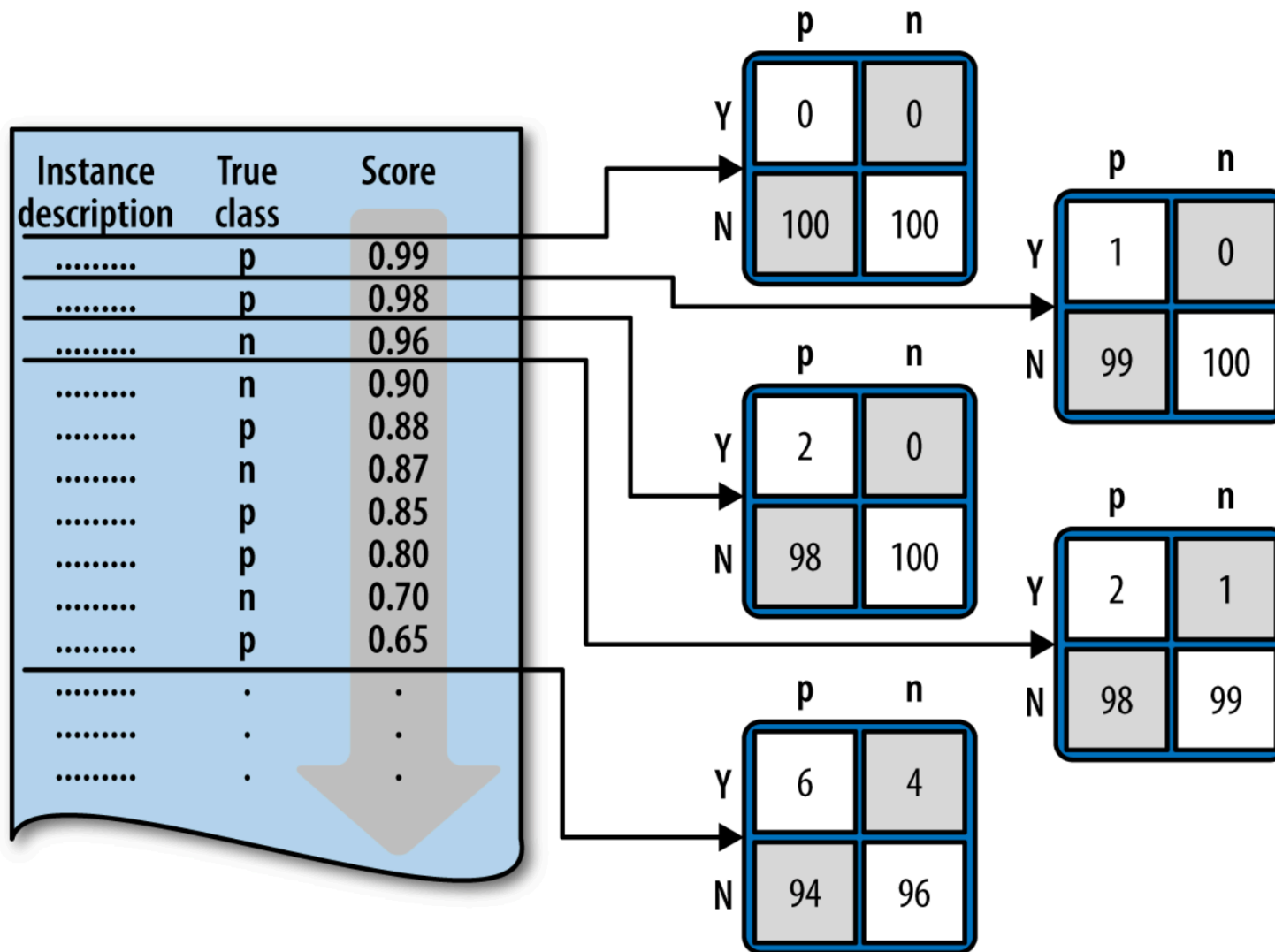
Profit Calculation

Combining information from the **Confusion matrix** and the **Cost-Benefit matrix** we can calculate **Expected Profit!**

		Actual	
		p	n
Predicted	Y	$b(Y,p)$	$c(Y,n)$
	N	$c(N,p)$	$b(N,n)$

$$\begin{aligned} E[Profit] &= P(Y, p) \cdot b(Y, p) + P(Y, n) \cdot c(Y, n) + \\ &\quad P(N, p) \cdot c(N, p) + P(N, n) \cdot b(N, n) \\ &= P(Y|p) \cdot P(p) \cdot b(Y, p) + P(Y|n) \cdot P(n) \cdot c(Y, n) + \\ &\quad P(N|p) \cdot P(p) \cdot c(N, p) + P(N|n) \cdot P(n) \cdot b(N, n) \\ &= P(p) \cdot [P(Y|p) \cdot b(Y, p) + P(N|p) \cdot c(N, p)] + \\ &\quad P(n) \cdot [P(Y|n) \cdot c(Y, n) + P(N|n) \cdot b(N, n)] \end{aligned}$$

Building the Profit Curve



Building the Profit Curve

For a given model f , each threshold value T gives a point on the Profit Curve

Model score is the threshold probability classifying $+$ vs $-$

- 1 Allow T to be the maximum score
- 2 $TP = 0, FP = 0$
- 3 Calculate $E[Profit]$
- 4 For each observation, i :
 - If $\hat{\pi}_i > T \rightarrow$ increment TP
 - Else \rightarrow increment FP
- 5 Add point (% Test Instances predicted Positive, $E[Profit]$) to the Profit Graph

Increment T from max-score to min-score, repeating steps 1-4

Profit Curves!

