

## COST CURVES

- Directly plot the misclassification costs / error (in terms of prior probs)
- Might be easier to interpret than ROC, especially in case of different misclassification costs or priors

- $f_1$  and  $f_2$  with intersecting ROC curves
- $f_2$  dominates first, then  $f_1$

**BUT:** Unclear for which thresholds, costs or class distrib  $f_2$  better than  $f_1$

Nathalie Japkowicz (2004): Evaluating Learning Algorithms : A Classification Perspective. (p. 125)

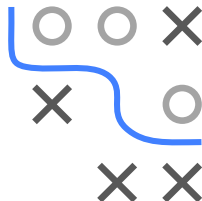
# COST CURVES

Simplifying assumption: equal misclassif costs, i.e.,  $cost_{FN} = cost_{FP}$

⇒ Expected misclassif cost reduces to misclassif error rate

With law of total prob, we write error rate as function of  $\pi_+$ :

$$\begin{aligned}\rho_{MCE}(\pi_+) &= (1 - \pi_+) \cdot \mathbb{P}(\hat{y} = 1 | y = 0) + \pi_+ \cdot \mathbb{P}(\hat{y} = 0 | y = 1) \\ &= (1 - \pi_+) \cdot FPR + \pi_+ \cdot FNR \\ &= (FNR - FPR) \cdot \pi_+ + FPR\end{aligned}$$



	Confusion matrix	
	True class	
	$y = 1$	$y = 0$
Pred. $\hat{y} = 1$	TP	FP
class $\hat{y} = 0$	FN	TN

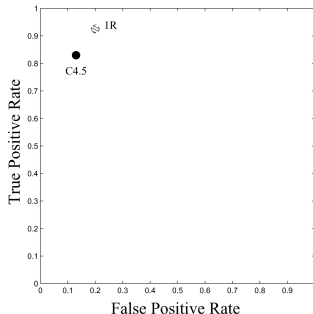
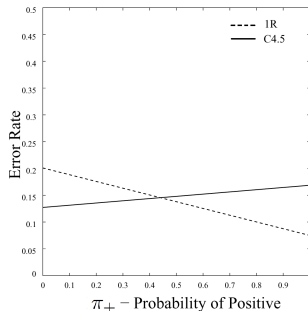
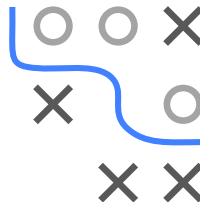
	Cost matrix	
	True class	
	$y = 1$	$y = 0$
Pred. $\hat{y} = 1$	0	$cost_{FP}$
class $\hat{y} = 0$	$cost_{FN}$	0

# COST CURVES

- Cost line of a classifier with slope ( $FNR - FPR$ ) and intercept  $FPR$ :

$$\rho_{MCE}(\pi_+) = (FNR - FPR) \cdot \pi_+ + FPR$$

- Cost curves are point-line duals of ROC curves, i.e., a single classifier is represented by a point in the ROC space and by a line in cost space



Chris Drummond and Robert C. Holte (2006): Cost curves: An improved method for visualizing classifier performance. Machine Learning, 65, 95-130 ([URL](#)).

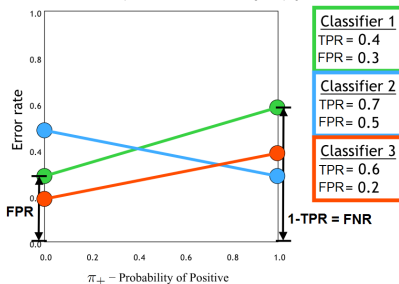
## COST LINES

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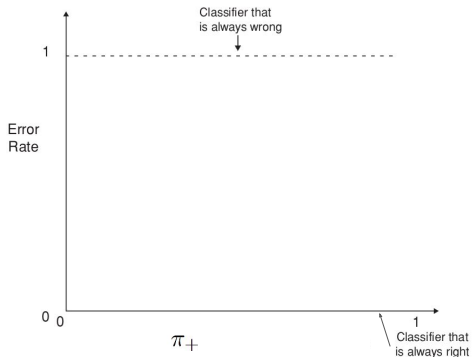
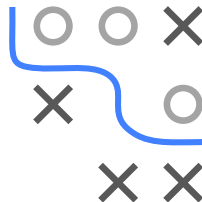
- Hard classifiers are points (TPR, FPR) in ROC space
- The cost line of a classifier connects  $(\pi_+, \rho_{MCE})$ -points at  $(0, FPR)$  and  $(1, 1 - TPR)$
- Classifier 3 always dominates classifier 1
- Classifier 3 is better than classifier 2 when  $\pi_+ < 0.7$

Cost lines plot different values of  $\pi_+$  vs.  $\rho_{MCE}(\pi_+)$



# COST LINES - EXAMPLE

- Horizontal dashed line: worst classifier (100% error rate for all  $\pi_+$ )  
 $\Rightarrow FNR = FPR = 1$
- x-axis: perfect classifier (0% error rate for all  $\pi_+$ )  $\Rightarrow FNR = FPR = 0$



$$\rho_{MCE} = (FNR - FPR) \cdot \pi_+ + FPR$$

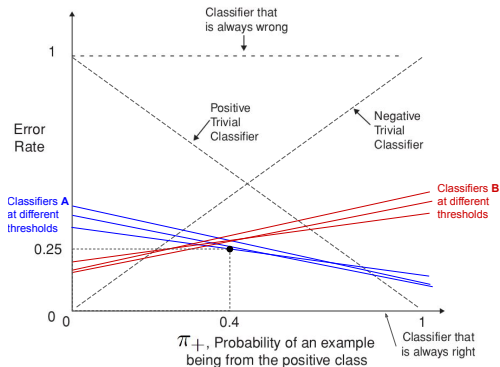
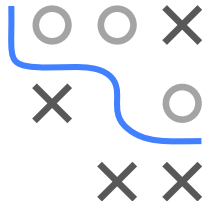
		Confusion matrix	
		True class	
Pred. class	$\hat{y} = 1$	TP	FP
	$\hat{y} = 0$	FN	TN

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- The graph illustrates the error rates of three classifiers as a function of  $\pi_+$  (the proportion of positive examples in the training set). The x-axis is labeled  $\pi_+$  and ranges from 0 to 1. The y-axis is labeled Error Rate and ranges from 0 to 1.
- Classifier that is always wrong:** A horizontal dashed line at Error Rate = 1.
  - Positive Trivial Classifier:** A dashed line from (0, 1) to (1, 0).
  - Negative Trivial Classifier:** A dashed line from (0, 0) to (1, 1).
- The intersection of the Positive and Negative Trivial Classifier lines is at  $\pi_+ = 0.5$  and Error Rate = 0.5.

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 $\Rightarrow FNR = FPR = 1$
- x-axis: perfect classifier (0% error rate for all  $\pi_+$ )  $\Rightarrow FNR = FPR = 0$
- Dashed diagonal lines: trivial classifiers, i.e., ascending diagonal always predicts negative instances ( $\Rightarrow FNR = 1$  and  $FPR = 0$ ) and vice versa
- Descending/ascending bold lines: two families of classifiers  $A$  and  $B$  (represented by points in their respective ROC curves)

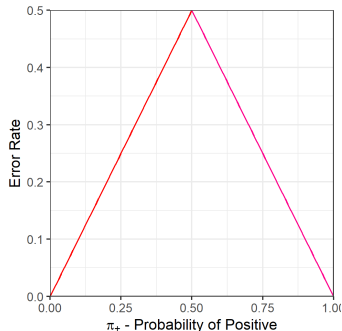
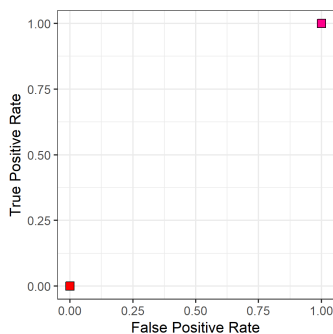


$$\rho_{MCE} = (FNR - FPR) \cdot \pi_+ + FPR$$

		Confusion matrix	
		True class	
Pred. class	$\hat{y} = 1$	$y = 1$	$y = 0$
	$\hat{y} = 0$	TP	FP
		FN	TN

# VISUALIZE COST CURVE - LOWER ENVELOPE

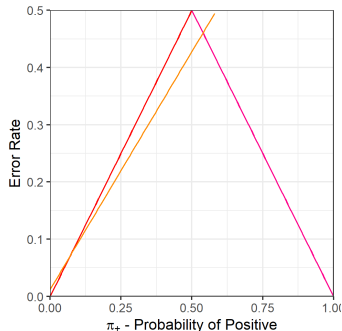
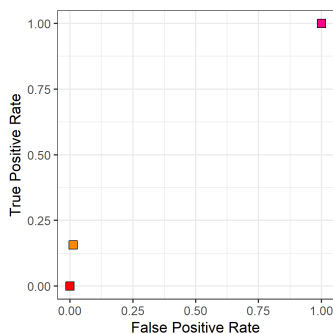
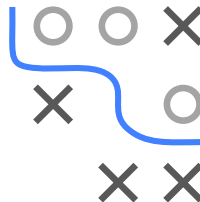
- Left: ROC = TPR & FPR of a classifier for different prob thresholds
- Right: Corresponding cost lines
- Duality: For every ROC point we can construct the CC line, and vice versa.





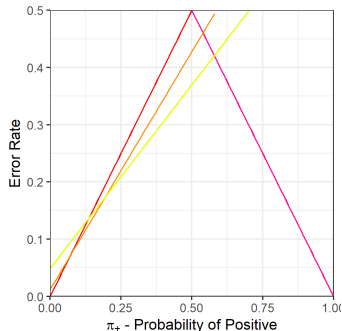
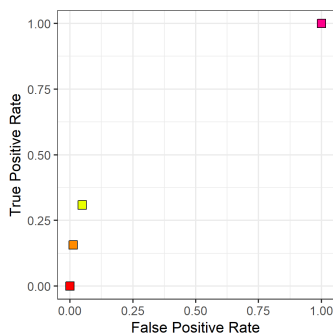
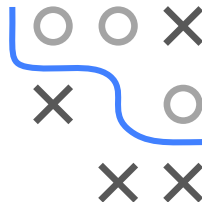
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# VISUALIZE COST CURVE - LOWER ENVELOPE

- Left: ROC = TPR & FPR of a classifier for different prob thresholds
- Right: Corresponding cost lines
- Duality: For every ROC point we can construct the CC line, and vice versa.
- **Cost curve** (right, black) is lower envelope of **cost lines**  
 $\triangleq$  pointwise minimum of error rate (as function of  $\pi_+$ )

