Cost Curves

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Joint work with Chris Drummond, NRC, Ottawa

Cost Curve Tool programmed by Alden Flatt



How to Evaluate Performance ?

- Scalar reasure ummarizing performance
 - Accuacy
 - Experted cos
 - Area under the LOC curve
- Performance Visualization Techniques

 ROC curve
 - Cost Curve



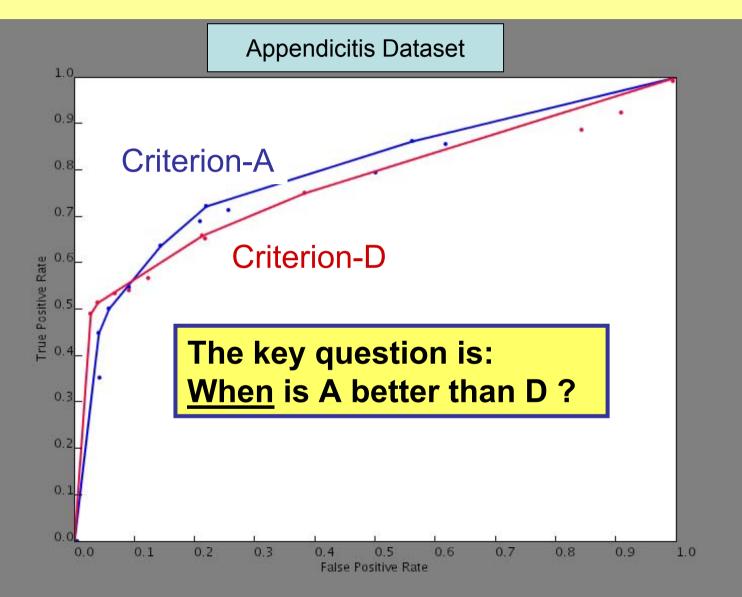
The Lure of Scalar Measures

- "...it is often preferable to employ a single value measure which summarizes the performance of a classifier, e.g. because there are several classifiers to be compared and there is no clear dominance of one ROC curve above the others.
 - The most widely used single measure is the Area Under the ROC Curve ..."

paraphrase from a workshop paper

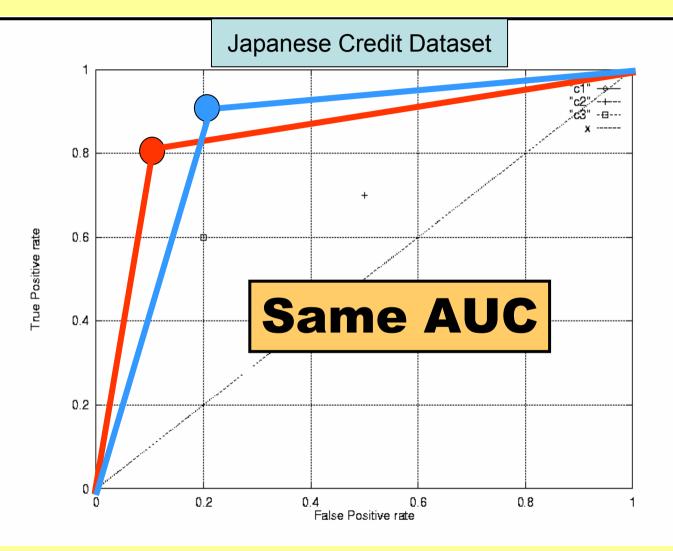


2 Splitting Criteria for C4.5



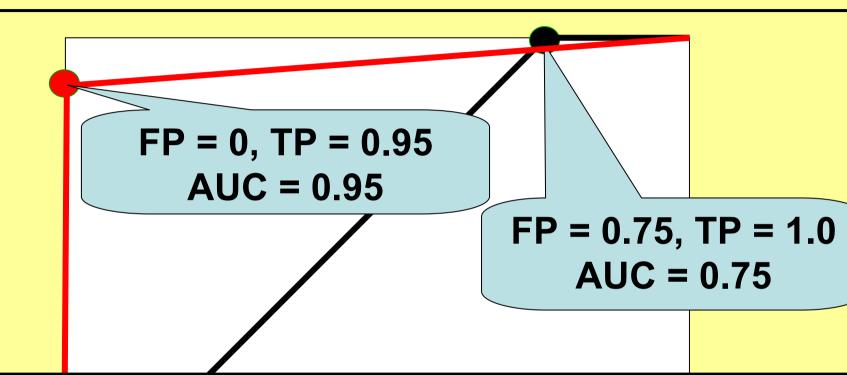
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C4.5 vs 1R





Is AUC=0.95 better than AUC=0.75?



When positives outnumber negatives 25:1, AUC=0.95 has more than twice the error rate of AUC=0.75*

* In Phil Long's application, the ratio is 327:1



What's Genuinely Good About Scalar Measures ?

 we know how to average them, compute confidence intervals, test for significance, etc.

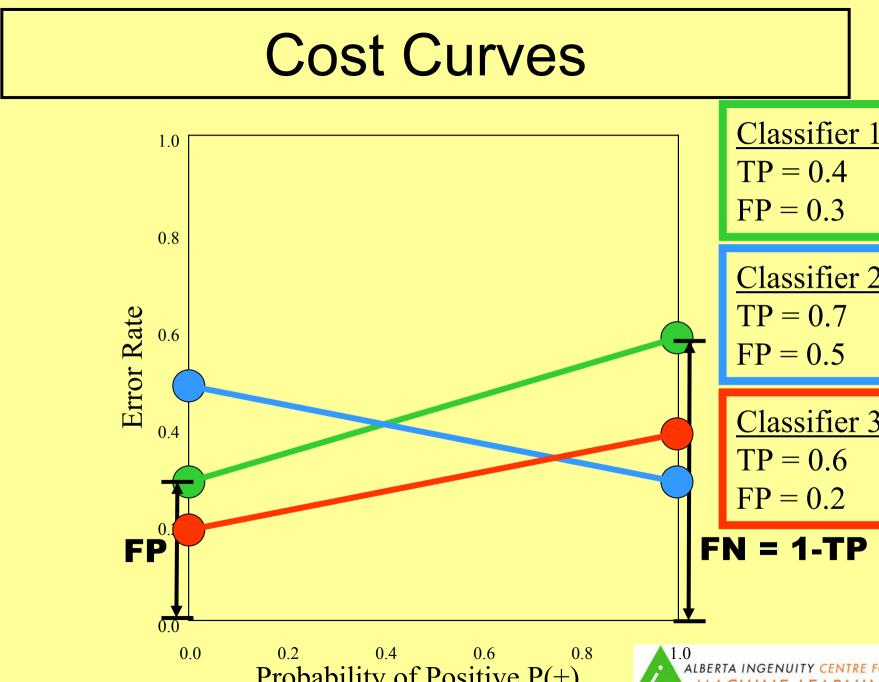
... and there is off-the-shelf software to do these calculations for us.

- being one-dimensional leaves the second dimension free for other uses, e.g.
 - Learning curves
 - Multiple datasets

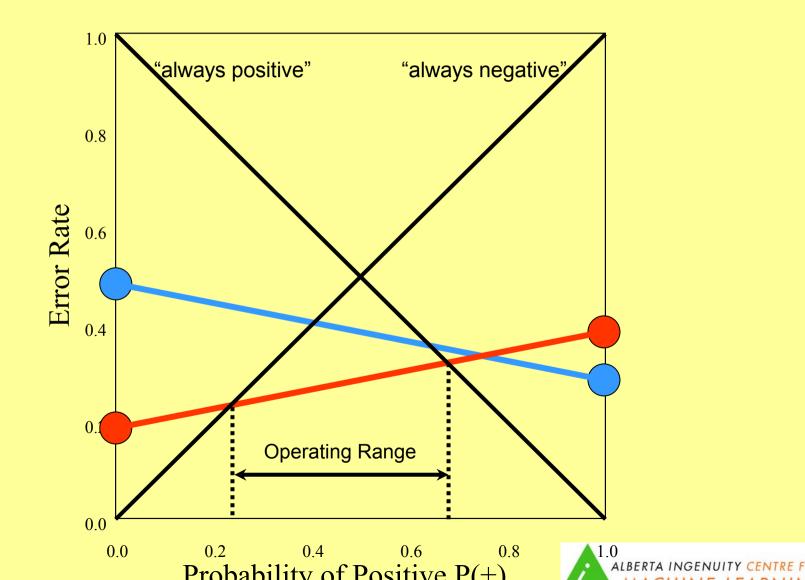


Cost Curves





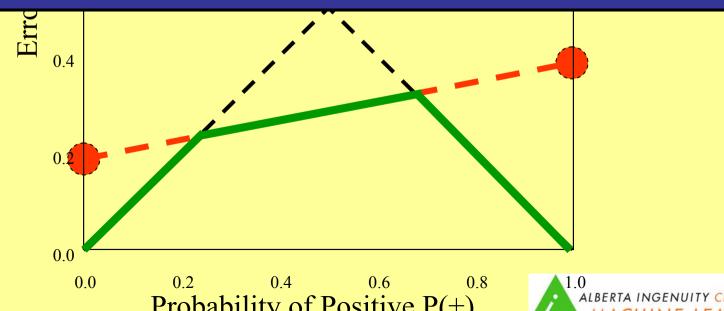
Operating Range



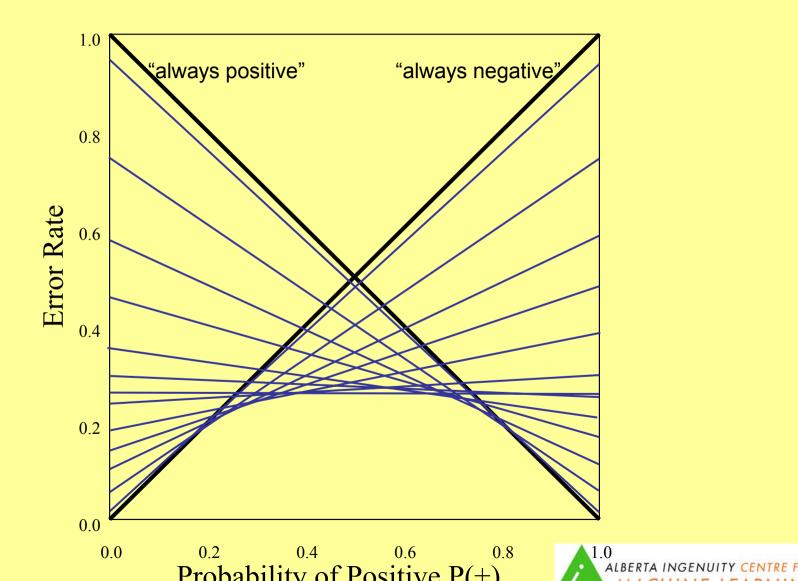
Lower Envelope



The lower envelope is a <u>biased estimate</u> of performance. Fresh data is needed to get an unbiased estimate.



Varying a Threshold



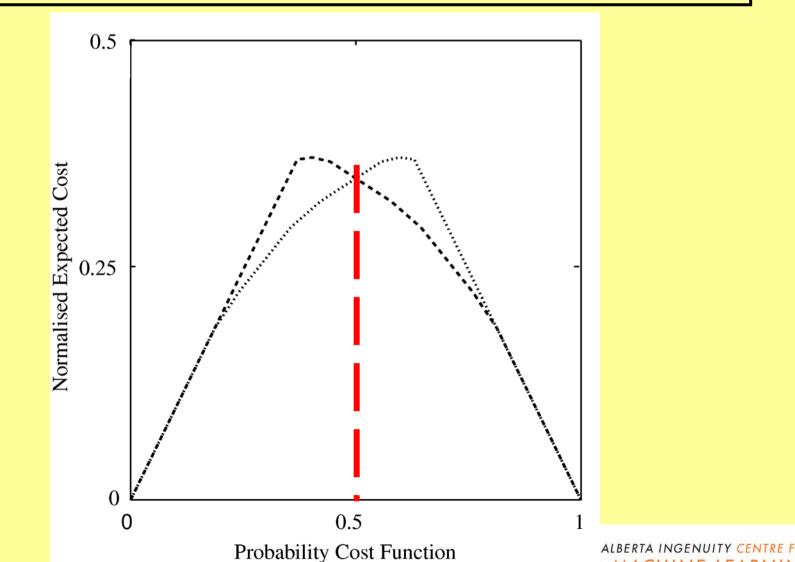
Taking Costs Into Account

$$X = \frac{p(+) \cdot C(-|+)}{p(+) \cdot C(-|+) + (1-p(+)) \cdot C(+|-)}$$

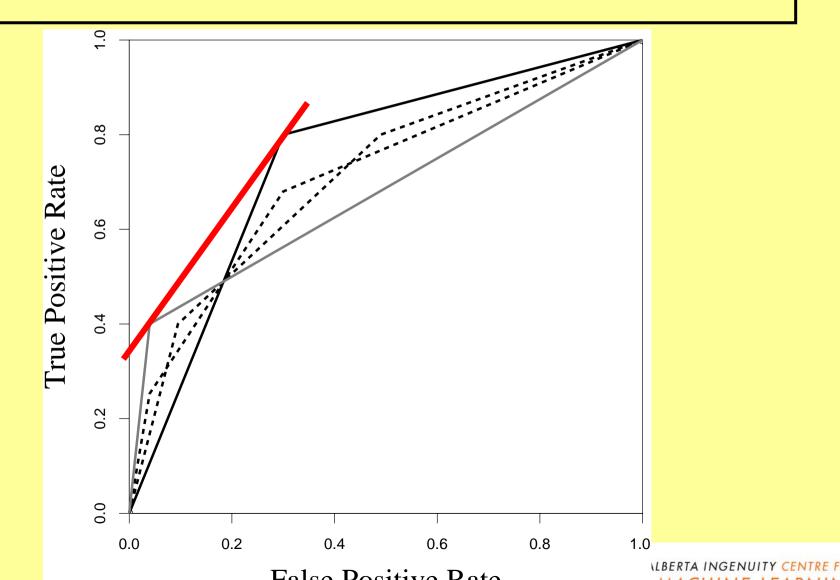
Y = expected cost normalized to [0,1]



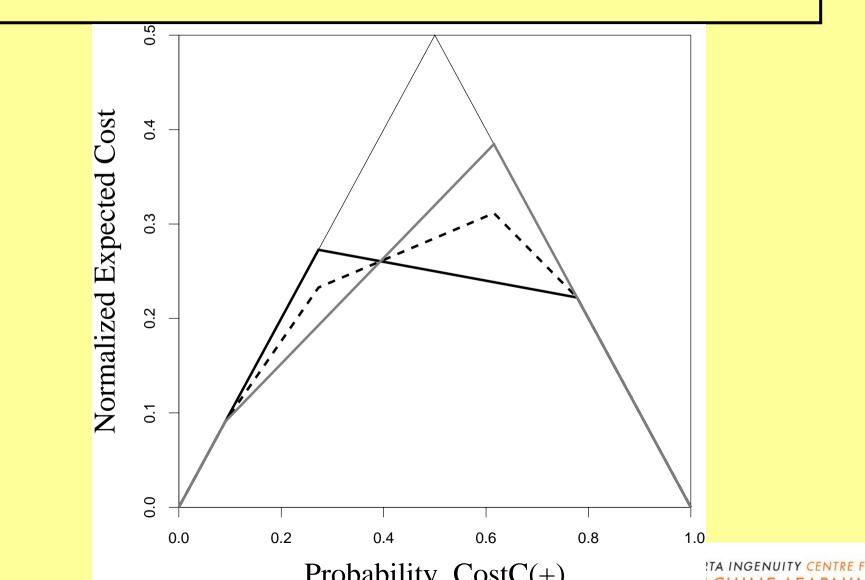
Comparing Cost Curves



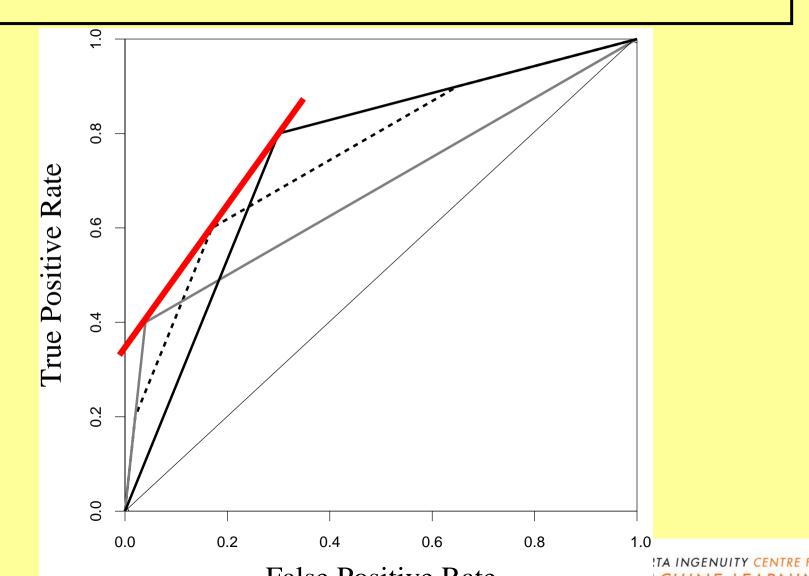
Averaging ROC Curves



Averaging Cost Curves



Cost Curve Avg. in ROC Space



Confidence Intervals

	Predicted	
True	pos	neg
pos	78	22
neg	40	60

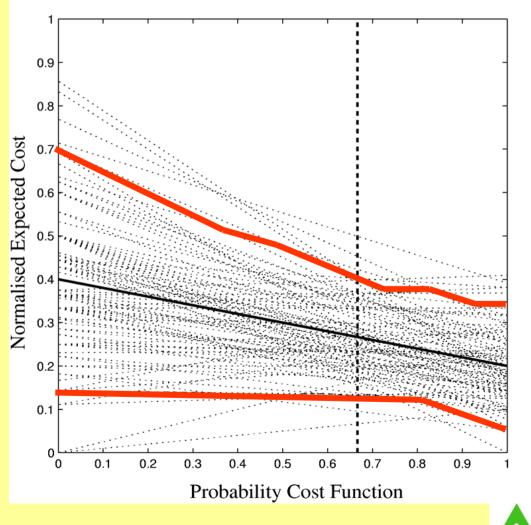
	Predicted	
True	pos	neg
pos	75	25
neg	45	55

	Predicted	
True	pos	neg
pos	83	17
neg	38	62

Resample confusion matrix 10000 times and take 95% envelope



Confidence Interval Example



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Paired Resampling to Test Statistical Significance

For the 100 test examples in the <u>negative</u> class:

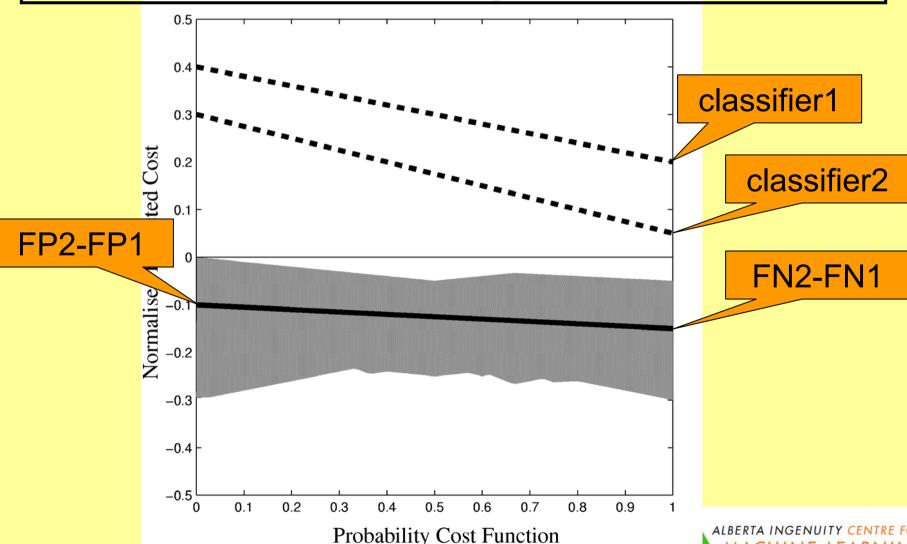
Predicted by	Predicted by Classifier2	
Classifier1	pos	neg
pos	30	10
neg	0	60

FP for classifier1: (30+10)/100 = 0.40FP for classifier2: (30+0)/100 = 0.30FP2 - FP1 = -0.10

Resample this matrix 10000 times to get (FP2-FP1) values. Do the same for the matrix based on positive test examples. Plot and take 95% envelope as before.



Paired Resampling to Test Statistical Significance



Correlation between Classifiers

High Correlation

Predicted by	Predicted by Classifier2	
Classifier1	pos	neg
pos	30	10
neg	0	60

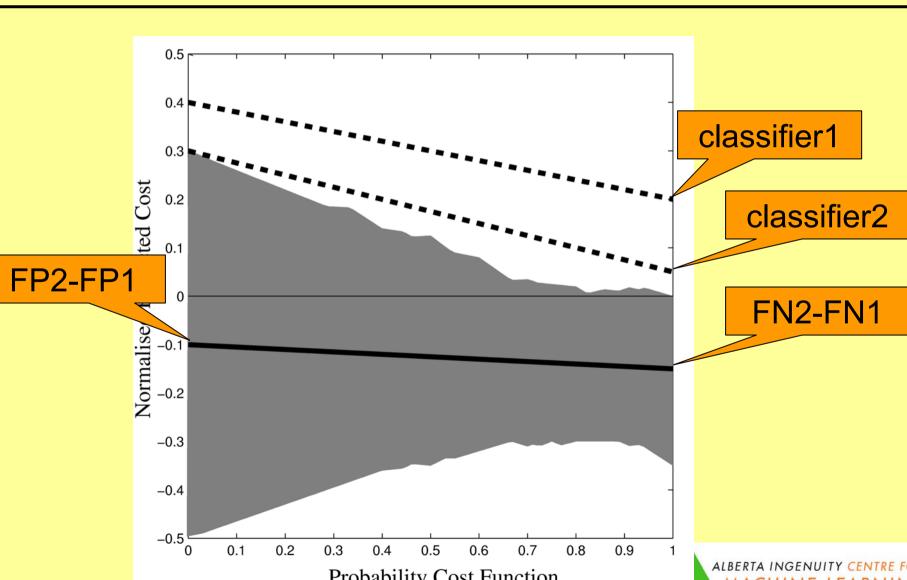
Low Correlation (same FP1 and FP2 as above)

Predicted by	Predicted by Classifier2	
Classifier1	pos	neg
pos	0	40
neg	30	30

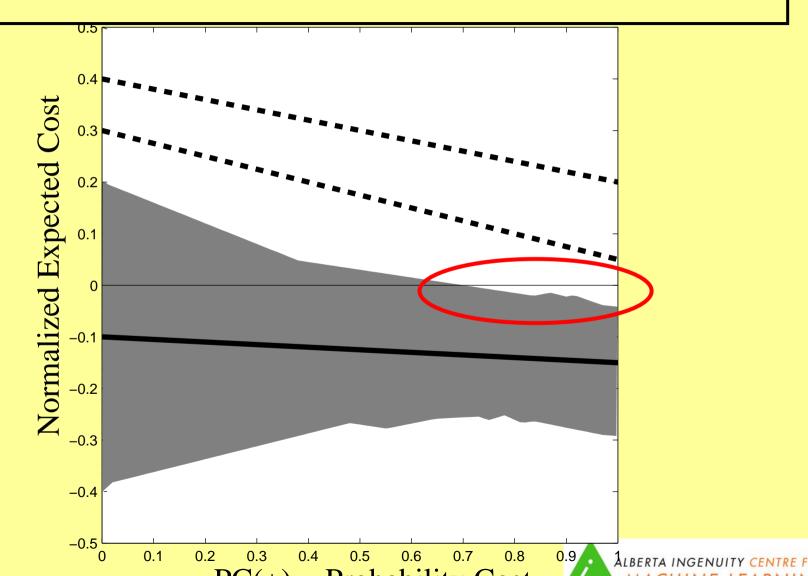


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Low correlation = Low significance



Limited Range of Significance



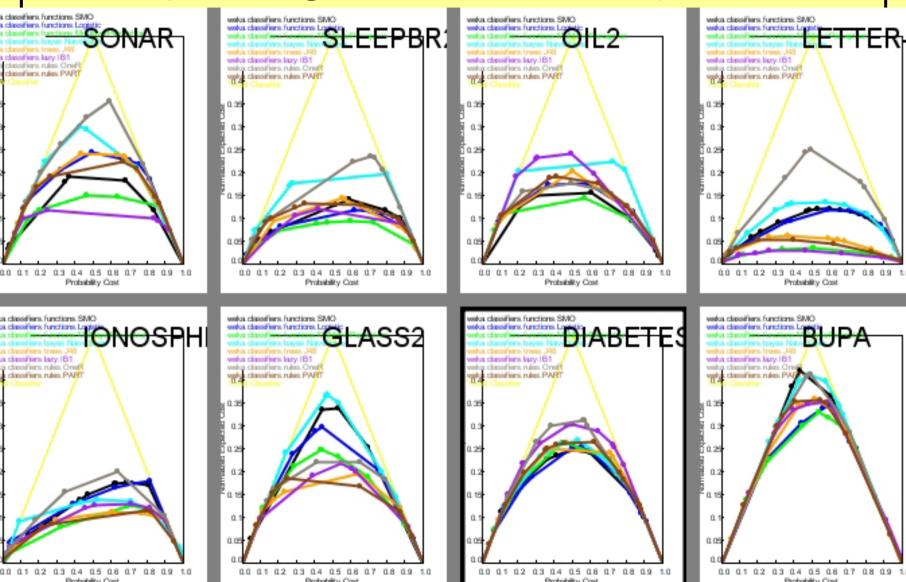
Comparing J48 and AdaBoost



Lower Envelope is Biased

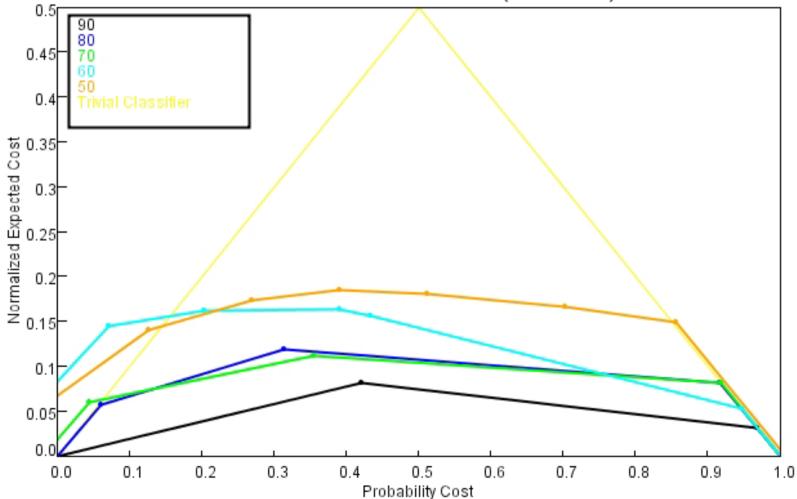


Multiple Alg/Dataset Comparisons



Learning Curves

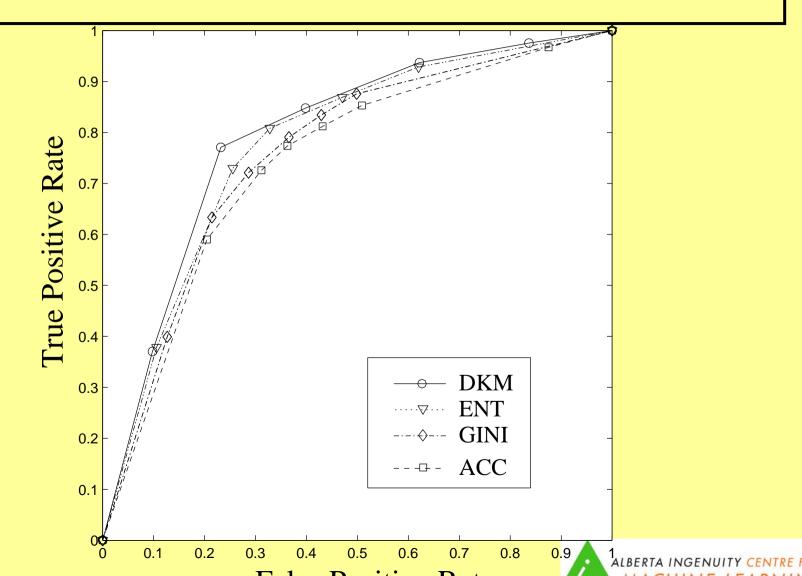
J48 Credit (seed=11)



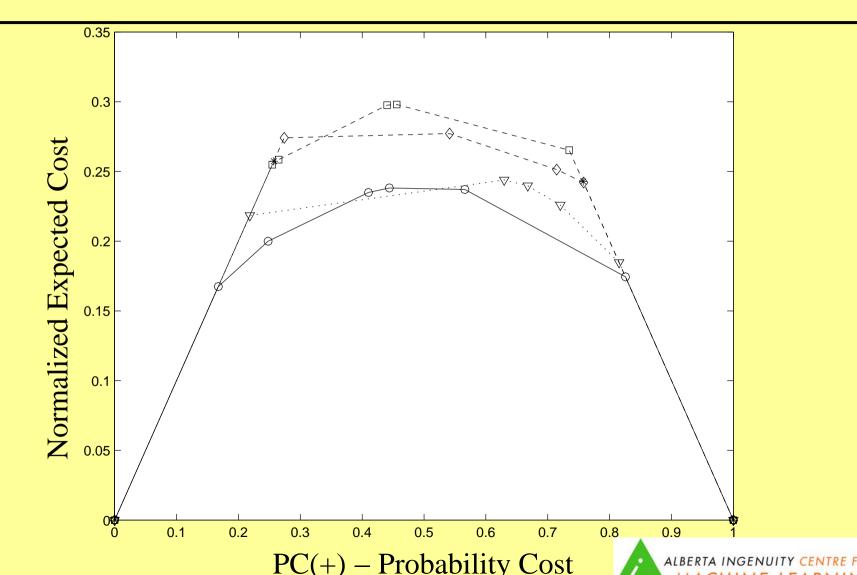
Better Data Analysis



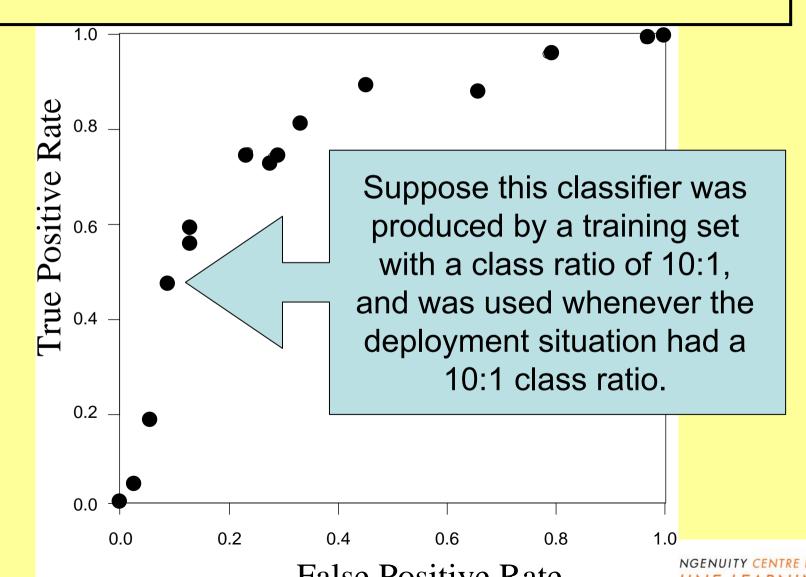
ROC, C4.5 Splitting Criteria



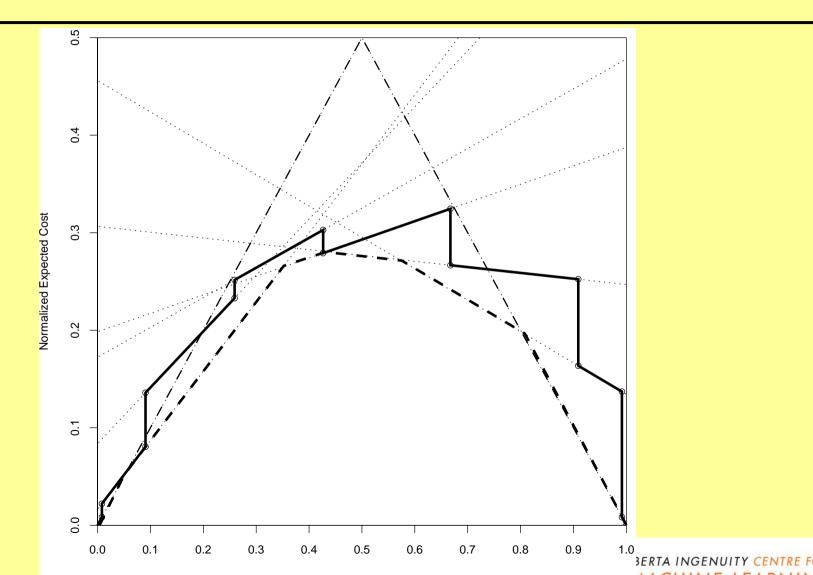
Cost Curve, C4.5 Splitting Criteria



ROC, Selection procedure



Cost Curves, Selection Procedure



Conclusions

- Scalar performance measures, including AUC, do not indicate <u>when</u> one classifier is better than another.
- Cost curves enable easy visualization of
 - Average performance (expected cost)
 - operating range
 - confidence intervals on performance
 - difference in performance and its significance
- Cost/ROC curve software is available. Contact: holte@cs.ualberta.ca

