

Evaluation – Counting  
the Costs

CSCI 347,  
Data Mining

# Confusion Matrix for Binary Classification

		Predicted	
		Yes	No
Actual	Positive	TP	FN
	Negative	FP	TN

On diagonal:

TP – true positive

TN – true negative

False results in corners

# The Kappa Statistic

*Kappa* statistic - measures relative improvement over random predictor

$$\frac{D_{observed} - D_{random}}{D_{perfect} - D_{random}}$$

Two confusion matrices for a 3-class problem:  
actual predictor (left) vs. random predictor (right)

		Predicted class						Predicted class			
		<i>a</i>	<i>b</i>	<i>c</i>	<i>total</i>			<i>a</i>	<i>b</i>	<i>c</i>	<i>total</i>
Actual class	<i>a</i>	88	10	2	100	Actual class	<i>a</i>	60	30	10	100
	<i>b</i>	14	40	6	60		<i>b</i>	36	18	6	60
	<i>c</i>	18	10	12	40		<i>c</i>	24	12	4	40
<i>total</i>		120	60	20		<i>total</i>		120	60	20	

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$$\frac{D_{observed} - D_{random}}{D_{perfect} - D_{random}}$$

$$\frac{140 - 82}{200 - 82} = \frac{58}{118} = 0.49|$$

# Cost-Sensitive Learning

Cost-sensitive learning –

- Can skew the data according to the costs
- Can weight instances according to the costs ,for example, give more weight to an instance that costs less (some versions of Naïve Bayes allow weighting the instances)

# Lift Charts

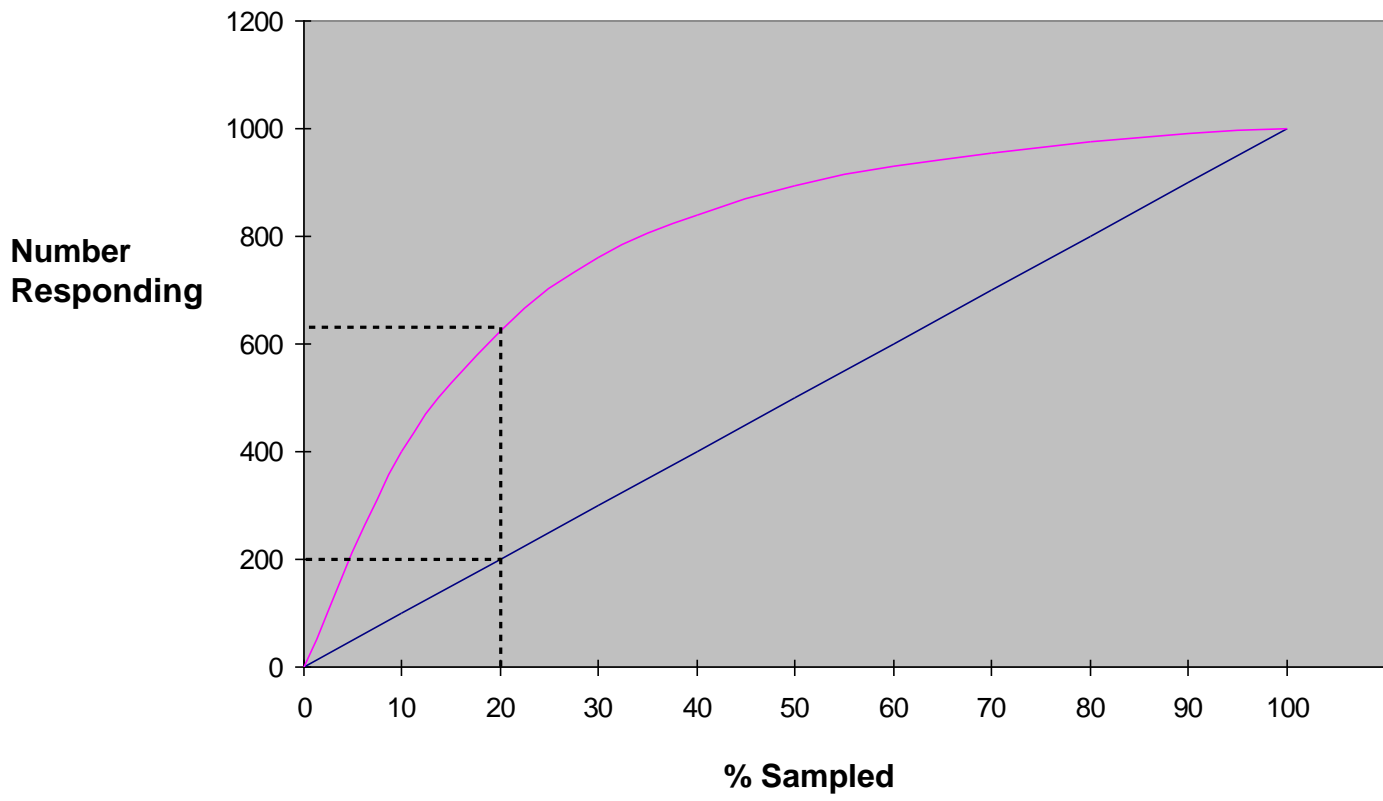
Example: promotional mailout to 1,000,000 households

Mail to all; 0.1% respond (1000)

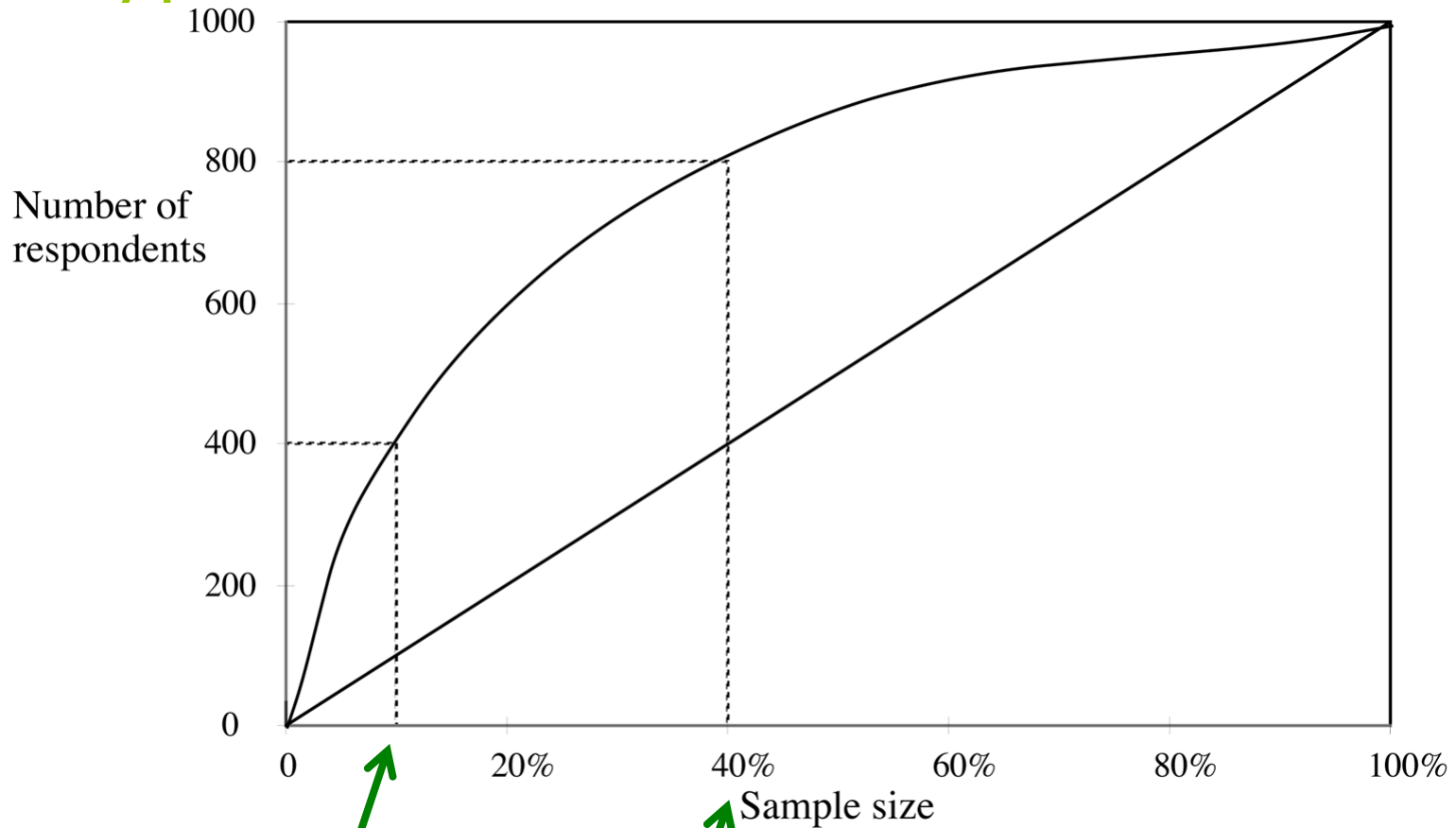
Data mining tool identifies subset of 100,000 most promising, 0.4% of these respond (400)  
*40% of responses for 10% of cost may pay off*

Identify subset of 400,000 most promising, 0.2% respond (800)

*A lift chart allows a visual comparison*



# A Hypothetical Lift Chart



40% of responses  
for 10% of cost

80% of responses  
for 40% of cost



# More Measures...

$$\textit{Precision} = TP / (TP + FP)$$

Test specificity

When the test predicts the person has the disease, how often is it true?

$$\textit{Recall} = TP / (TP + FN)$$

Test sensitivity

When a person has the disease, how often does the test find it?

Summary measure:

$$\textit{F-measure} = (2 \times \textit{recall} \times \textit{precision}) / (\textit{recall} + \textit{precision})$$