## Subgroup Discovery

## Exploratory Data Analysis

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Q: How can we use ideas from C4.5 to approach this task?

A: Why not list the info gain of all attributes, and rank according to this?

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## Interactions between Attributes

- Single-attribute effects are not enough
- XOR problem is extreme example: 2 attributes with no info gain form a good model
- Apart from
$A=\mathrm{a}, B=\mathrm{b}, C=\mathrm{c}, \ldots$
- consider also
$A=\mathrm{a} \wedge B=\mathrm{b}, A=\mathrm{a} \wedge C=\mathrm{c}, \ldots, B=\mathrm{b} \wedge C=\mathrm{c}, \ldots$
$A=\mathrm{a} \wedge B=\mathrm{b} \wedge C=\mathrm{c}, \ldots$


## Subgroup Discovery Task

"Find all subgroups within the inductive constraints that show a significant deviation in the distribution of the target attribute"

- Inductive constraints:
- Minimum support
- (Maximum support)
- Minimum quality (Information gain, $X^{2}$, WRAcc)
- Maximum complexity


## Confusion Matrix

- A confusion matrix (or contingency table) describes the frequency of the four combinations of subgroup and target:
" within subgroup, positive
- within subgroup, negative
" outside subgroup, positive



## Confusion Matrix

- High numbers along the TT-FF diagonal means a positive correlation between subgroup and target
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## Quality Measures

A quality measure for subgroups summarizes the interestingness of its confusion matrix into a single number

WRAcc, weighted relative accuracy

- $\operatorname{WRAcc}(\mathrm{S}, \mathrm{T})=p(\mathrm{ST})-p(\mathrm{~S}) \cdot p(\mathrm{~T})$
- between -. 25 and .25, 0 means uninteresting
- Balance between coverage and unexpectedness



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## Quality Measures

- WRAcc: Weighted Relative Accuracy
- Information gain
- $X^{2}$
- Correlation Coefficient
- Laplace
- Jaccard
- Specificity


## Subgroup Discovery as Search

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Refinements are (anti-) monotonic in their support...

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Refinements are (anti-) monotonic in their support...
...but not in interestingness. This may go up or down.

## Subgroup Discovery and ROC space



## ROC Space

ROC $=$ Receiver Operating Characteristics


Each subgroup forms a point in ROC space, in terms of its False Positive Rate, and True Positive Rate.

TPR = TP/Pos = TP/TP+FN (fraction of positive cases in the subgroup) FPR $=\mathrm{FP} / \mathrm{Neg}=\mathrm{FP} / \mathrm{FP}+\mathrm{TN}$ (fraction of negative cases in the subgroup)

## ROC Space Properties



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## Measures in ROC Space




## Measures in ROC Space



isometric

## Other Measures

Precision


Correlation coefficient


Gini index


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## Refinements in ROC Space



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Blue polygon represents possible refinements of $S$. With a convex measure, $f$ is bounded by measure of corners.

If corners are not above minimum quality or current best (top $k$ ?), prune search space below S .

## Combining Two Subgroups



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## Multi-class problems

- Generalising to problems with more than 2 classes is fairly staightforward:



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## Numeric Subgroup Discovery



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- Target is numeric: find subgroups with significantly higher or lower average value



## Numeric Subgroup Discovery

- Target is numeric: find subgroups with significantly higher or lower average value
- Trade-off between size of subgroup and average target value

Quiz 1

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