

How far can we go?

```
IF pclass='1' AND sex='female' THEN survive=yes
IF pclass='2' AND sex='female' THEN survive=yes
IF pclass='3' AND sex='female' AND age < 4 THEN survive=yes
IF pclass='3' AND sex='female' AND age >= 4 THEN survive=no
IF pclass='2' AND sex='male' THEN survive=no
IF pclass='3' AND sex='male' THEN survive=no
IF pclass='1' AND sex='male' AND age < 5 THEN survive=yes
...
```

Sequential Covering

```
Initialize R to the empty set
for each class C {
   while D is nonempty {
       Construct one rule r that correctly classifies
       some instances in D that belong to class C and
       does not incorrectly classify any non-C instances
       Add rule r to ruleset R
       Remove from D all instances correctly classified by r
return R
```

Bill Howe, UW

Sequential Covering: Finding next rule for class C

```
Initialize A as the set of all attributes over D
while r incorrectly classifies some non-C instances of D {
    write r as ant(r) \Rightarrow C
    for each attribute-value pair (a=v),
    where a belongs to A and v is a value of a,
    compute the accuracy of the rule
         ant(r) and (a=v) => C
    let (a^*=v^*) be the attribute-value pair of
    maximum accuracy over D; in case of a tie,
    choose the pair that covers the greatest
    number of instances of D
    update r by adding (a*=v*) to its antecedent:
        r = (ant(r) and (a*=v*)) => C
    remove the attribute a* from the set A:
        A = A - \{a^*\}
```

Strategies for Learning Each Rule

- General-to-Specific
 - Start with an empty rule
 - Add constraints to eliminate negative examples
 - Stop when only positives are covered
- Specific-to-General (not shown)
 - Start with a rule that identifies a single random instance
 - Remove constraints in order to cover more positives
 - Stop when further generalization results in covering negatives

Conflicts

- If more than one rule is triggered
 - Choose the "most specific" rule
 - Use domain knowledge to order rules by priority

Recap

- Representation
 - A set of rules: IF...THEN conditions
- Evaluation
 - coverage: # of data points that satisfy conditions
 - accuracy = # of correct predictions / coverage
- Optimization
 - Build rules by finding conditions that maximize accuracy

 One rule is easy to

interpret, but a complex set of rules probably isn't