

**Algorithm: Generate\_decision\_tree.** Generate a decision tree from the training tuples of data partition,  $D$ .

**Input:**

- Data partition,  $D$ , which is a set of training tuples and their associated class labels;
- *attribute\_list*, the set of candidate attributes;
- *Attribute\_selection\_method*, a procedure to determine the splitting criterion that “best” partitions the data tuples into individual classes. This criterion consists of a *splitting\_attribute* and, possibly, either a *split-point* or *splitting subset*.

**Output:** A decision tree.

**Method:**

- (1) create a node  $N$ ;
- (2) **if** tuples in  $D$  are all of the same class,  $C$ , **then**
- (3)     return  $N$  as a leaf node labeled with the class  $C$ ;
- (4) **if** *attribute\_list* is empty **then**
- (5)     return  $N$  as a leaf node labeled with the majority class in  $D$ ; // majority voting
- (6) apply **Attribute\_selection\_method**( $D$ , *attribute\_list*) to **find** the “best” *splitting\_criterion*;
- (7) label node  $N$  with *splitting\_criterion*;
- (8) **if** *splitting\_attribute* is discrete-valued **and**  
      multiway splits allowed **then** // not restricted to binary trees
- (9)     *attribute\_list*  $\leftarrow$  *attribute\_list*  $-$  *splitting\_attribute*; // remove *splitting\_attribute*
- (10) **for each** outcome  $j$  of *splitting\_criterion*  
      // partition the tuples and grow subtrees for each partition
- (11)     let  $D_j$  be the set of data tuples in  $D$  satisfying outcome  $j$ ; // a partition
- (12)     **if**  $D_j$  is empty **then**
- (13)         attach a leaf labeled with the majority class in  $D$  to node  $N$ ;
- (14)     **else** attach the node returned by **Generate\_decision\_tree**( $D_j$ , *attribute\_list*) to node  $N$ ;
- endfor**
- (15) return  $N$ ;