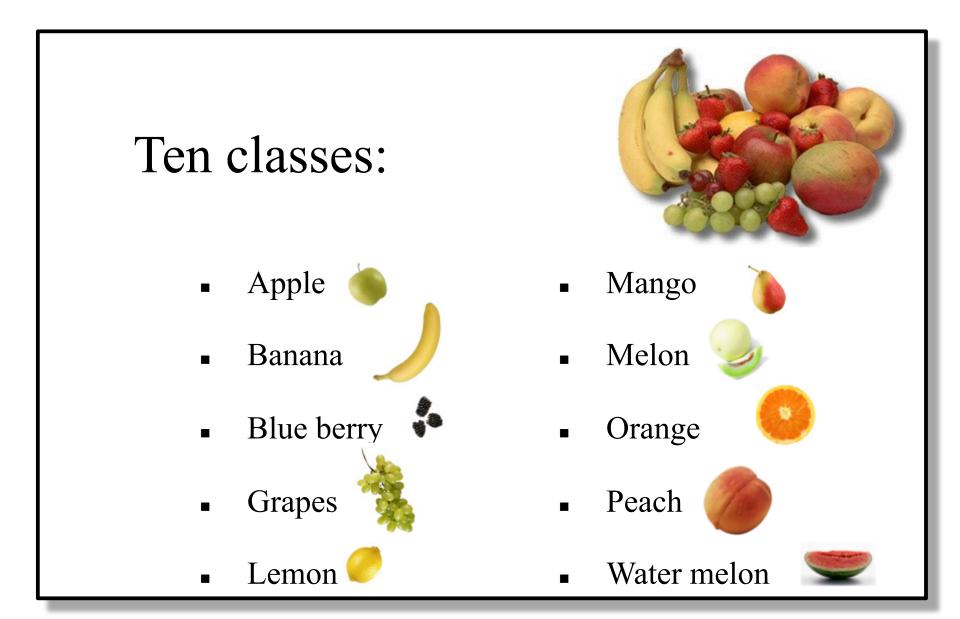
#### **Decision Tree**

#### Example – fruit classification





#### Four features:

Color: [yellow; green; red; blue; orange]

■ Size: [xtra-small; small; medium

large ; extra-large]

■ Shape: [<u>ro</u>und ; <u>e</u>lipsoidal ; <u>n</u>arrow

round-with-concavity]

■ Texture: [smooth; citrus]

#### Feature Vectors:



<color, size, shape, texture>

#### Examples:



<y,m,n,s>



<b,xs,ro,s>



<g,sm,ro,s>



<y,m,ro,c>

#### Given a set S with 25 labeled patterns:



<r,m,rwc,s>; <g,m,rwc,s> <y,m,rwc,s>



<g,m,e,s>; <y,m,e,s>



<y,m,n,s>; <g,m,n,s> <y,l,n,s>



<y,l,ro,s>; <y,l,e,s> <g,l,e,s>



<b,xs,ro,s> ; <b,xs,ro,s>



<o,m,ro,c>; <o,m,ro,c>



<b,sm,ro,s> ; <g,sm,ro,s> <y,sm,ro,s>

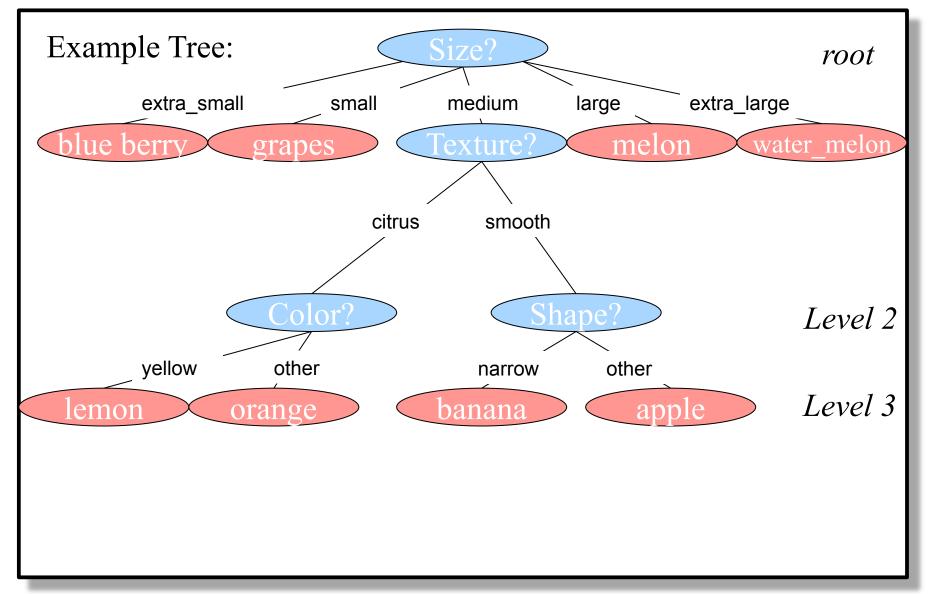


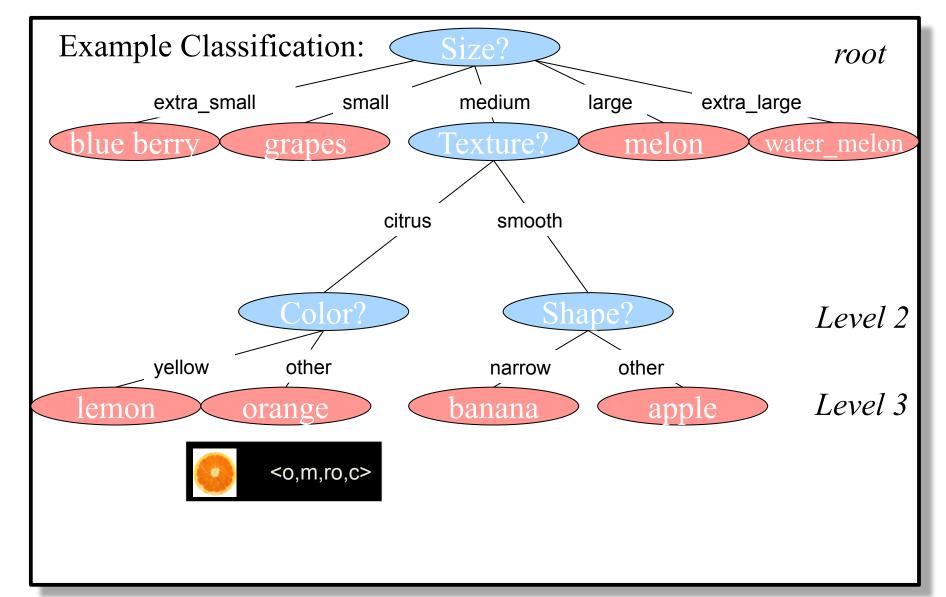
<y,m,rwc,s>; <r,m,rwc,s>

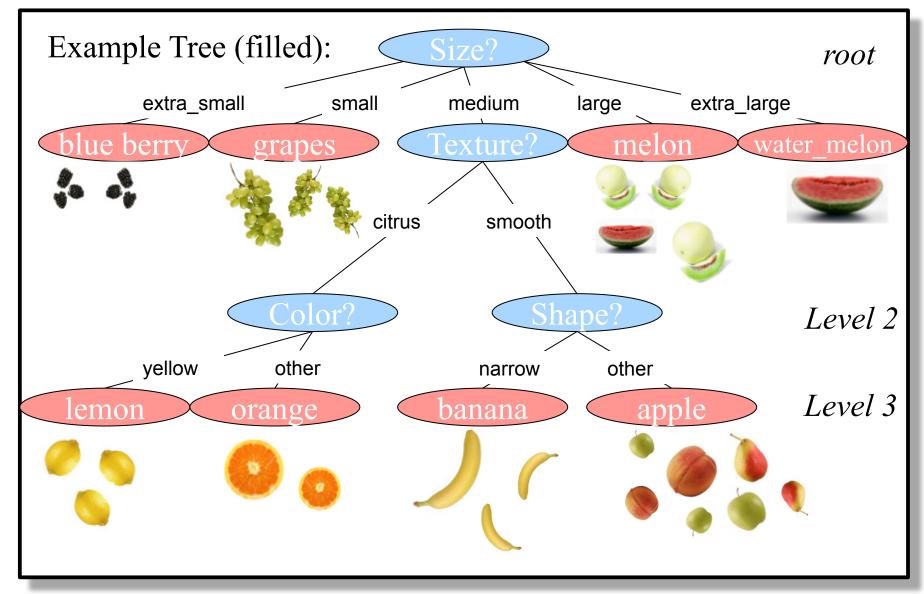


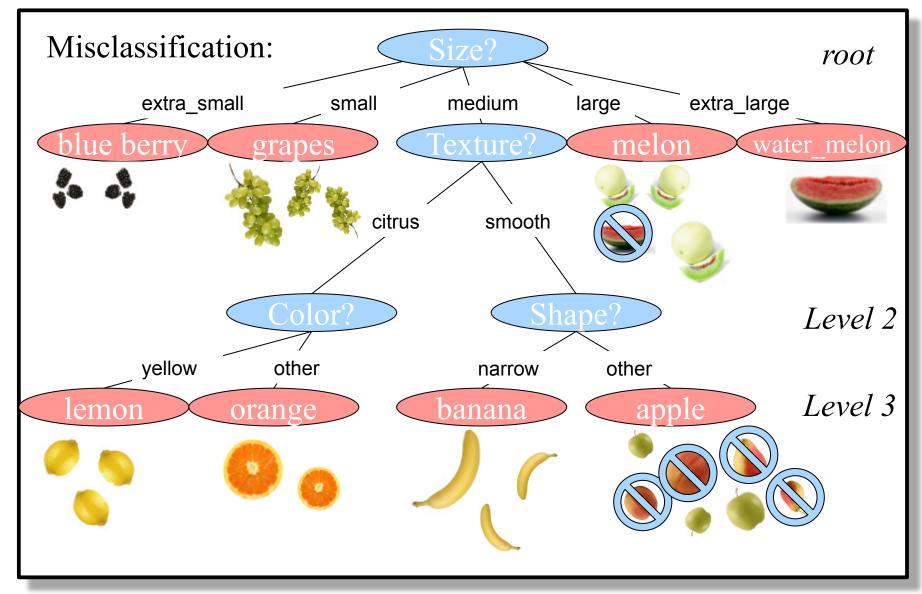
<y,m,ro,c> ; <g,m,ro,c> <y,m,e,c> <g,xl,ro,s> ; <g,l,ro,s>











#### Design of a decision tree

Problem: Given a training data set, how to create a decision tree?

General approach: split the set using a given feature, split the subsets using other features, ..., until you get pure (i.e. single class) or nearly pure subsets

#### Query selection and node impurity

The query at a node is selected to minimize the impurity of the two subsets descending from the node.

# **Impurity**

#### Misclassification impurity:

$$i(N) = 1 - \max_{j} P(\omega_{j})$$

The impurity of set S:



$$i(N) = 1 - \max_{j} P(\omega_{j})$$

$$i_{\rm S} = 1 - (3/25) = 22/25$$

### Query design

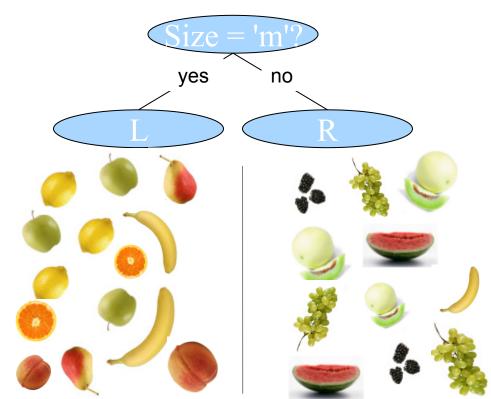
#### Impurity drop:

$$\Delta i(N) = i(N) - P_L i(N_L) - (1 - P_L) i(N_R)$$

 $N_L$  and  $N_R$  - left and right descendent nodes  $P_L$  - fraction of patterns which go to  $N_L$ 

Choose a query that maximizes the impurity drop

Suppose we create a tree with the following question: "Put an object in L if (size = m), otherwise put it in R" What is the impurity drop in this case?



$$i_{\rm S} = 22/25$$

$$i_{\rm L} = 1 - 3/14 = 11/14$$
  
 $i_{\rm R} = 1 - 3/11 = 8/11$ 

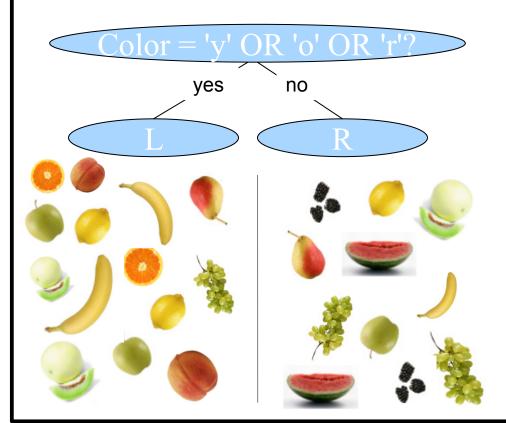
drop = 
$$i_{S} - P_{L} * i_{L} - (1 - P_{L}) * i_{R}$$

drop = 
$$22/25 - (14/25)*(11/14) - (1-14/25)*(8/11)$$

$$drop = 3/25$$

Suppose we chose a different question:

"Put an object in L if  $(colour = y \ OR \ o \ OR \ r)$ , otherwise put it in R" What is the impurity drop in this case?



$$i_{\rm S} = 22/25$$

$$i_{\rm L} = 1 - 2/14 = 12/14$$
  
 $i_{\rm R} = 1 - 2/11 = 9/11$ 

drop = 
$$i_S - P_L * i_L - (1 - P_L) * i_R$$

drop = 
$$22/25 - (14/25)*(12/14) - (1-14/25)*(9/11)$$

$$drop = 1/25$$

#### Fruit Classification

Which question was better?



"Put an object in L if (size = m), otherwise put it in R"

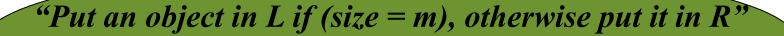
(impurity drop: 3/25)

"Put an object in L if  $(color = y \ OR \ o \ OR \ r)$ , otherwise put it in R"

(impurity drop: 1/25)

#### Fruit Classification

Which question was better?



impurity drop: 3/25

The higher the impurity drop, the better!

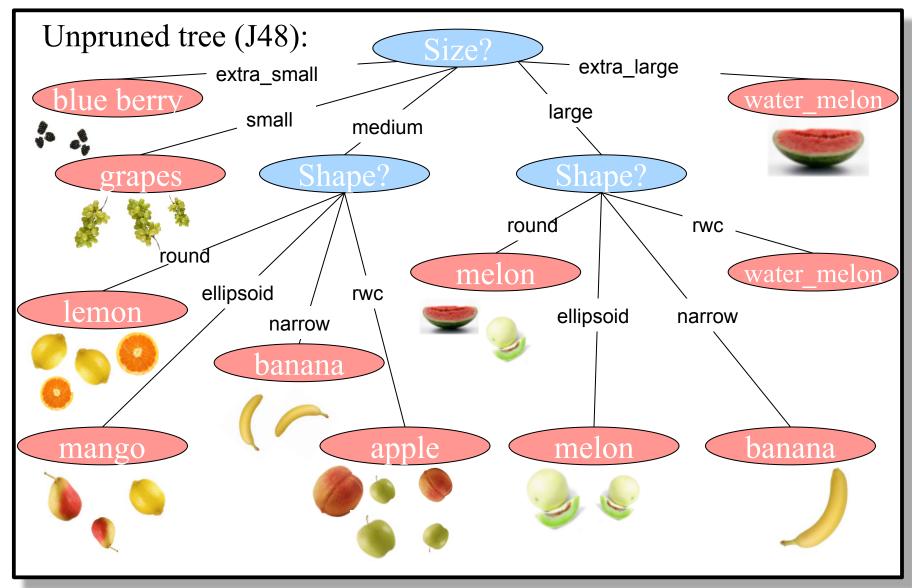
### Pruning

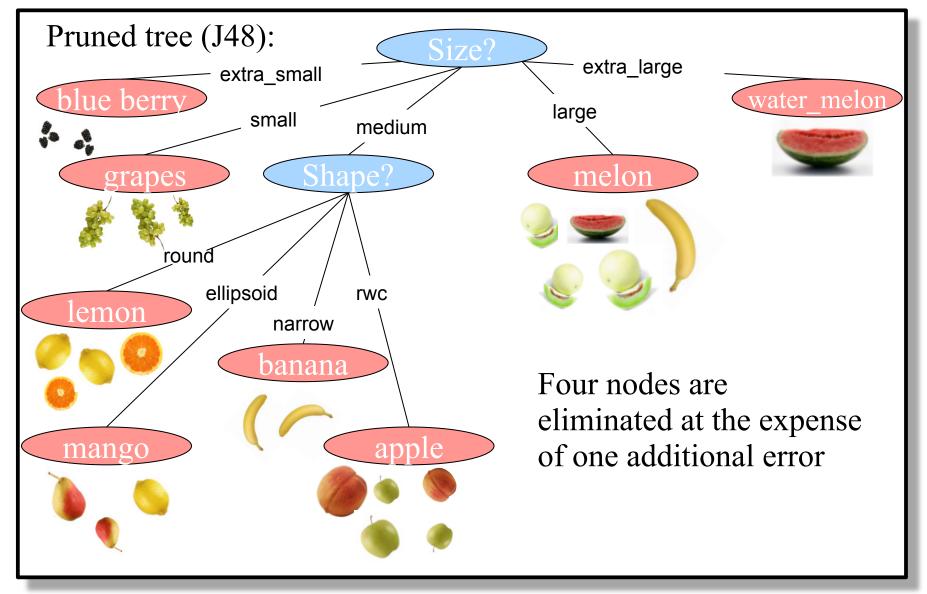
Early splitting can cut off the possibility of beneficial splits in descendant nodes (limited horizon effect)

Alternative to splitting: the tree is first fully grown, then

#### Pruning:

- Remove pairs of leaf nodes with a common parent or
- Replace a subtree by a leaf node.





# Summary of concepts

- Node impurity
- Query design
- Pruning