

k-Nearest Neighbors

CS 540

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K-nearest neighbors for classification

- Given training data $\{(x_i, y_i): 1 \leq i \leq n\}$
- Store the training data
- Given a new data point x^* , find its k nearest neighbors in the training data, predict the majority label of the neighbors

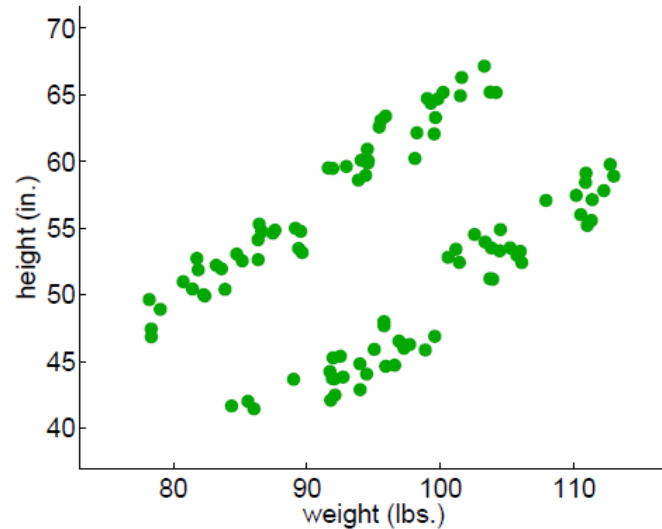
K-nearest neighbors for classification

*Input: Training data $(\mathbf{x}_1, y_1), \dots, (\mathbf{x}_n, y_n)$; distance function $d()$;
number of neighbors k ; test instance \mathbf{x}^**

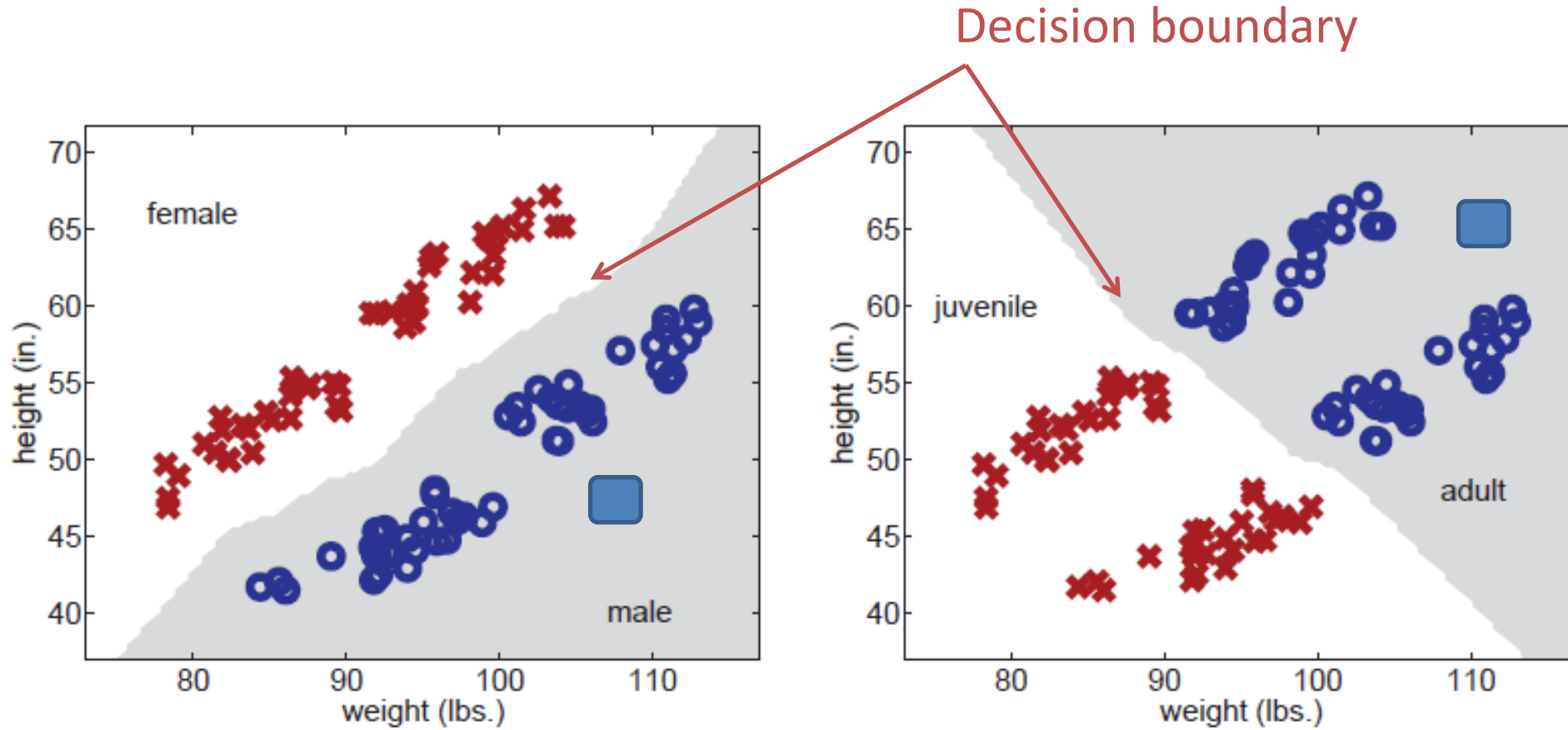
- 1. Find the k training instances $\mathbf{x}_{i_1}, \dots, \mathbf{x}_{i_k}$ closest to \mathbf{x}^* under distance $d()$.*
- 2. Output y^* as the majority class of y_{i_1}, \dots, y_{i_k} . Break ties randomly.*

Example: 1-NN for Little Green Man

- Little green men:
 - Predict gender (M,F) from weight, height?
 - Predict age (adult, juvenile) from weight, height?



Example: 1-NN for Little Green Man

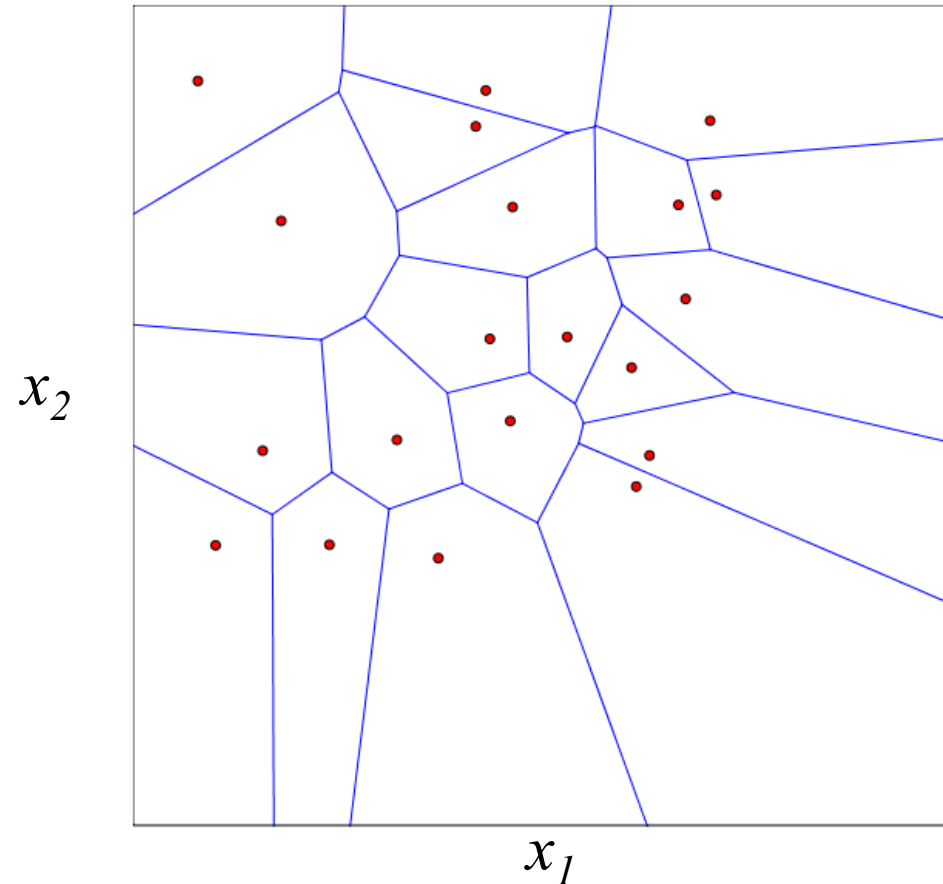


(a) classification by gender

(b) classification by age

The decision regions for 1-NN

Voronoi diagram: each polyhedron indicates the region of feature space that is in the nearest neighborhood of each training instance



K-NN for regression

- What if we want regression?
- Instead of majority vote, take average of neighbors' labels
 - Given test point x^* , find its k nearest neighbors x_{i_1}, \dots, x_{i_k}
 - Output the predicted label $\frac{1}{k} (y_{i_1} + \dots + y_{i_k})$

How can we determine distance

suppose all features are discrete

- Hamming distance: count the number of features for which two instances differ

suppose all features are continuous

- Euclidean distance: sum of squared differences

$$d(x_i, x_j) = \sum_f (x_{if} - x_{jf})^2, \text{ where } x_{if} \text{ is the } f\text{-th feature}$$

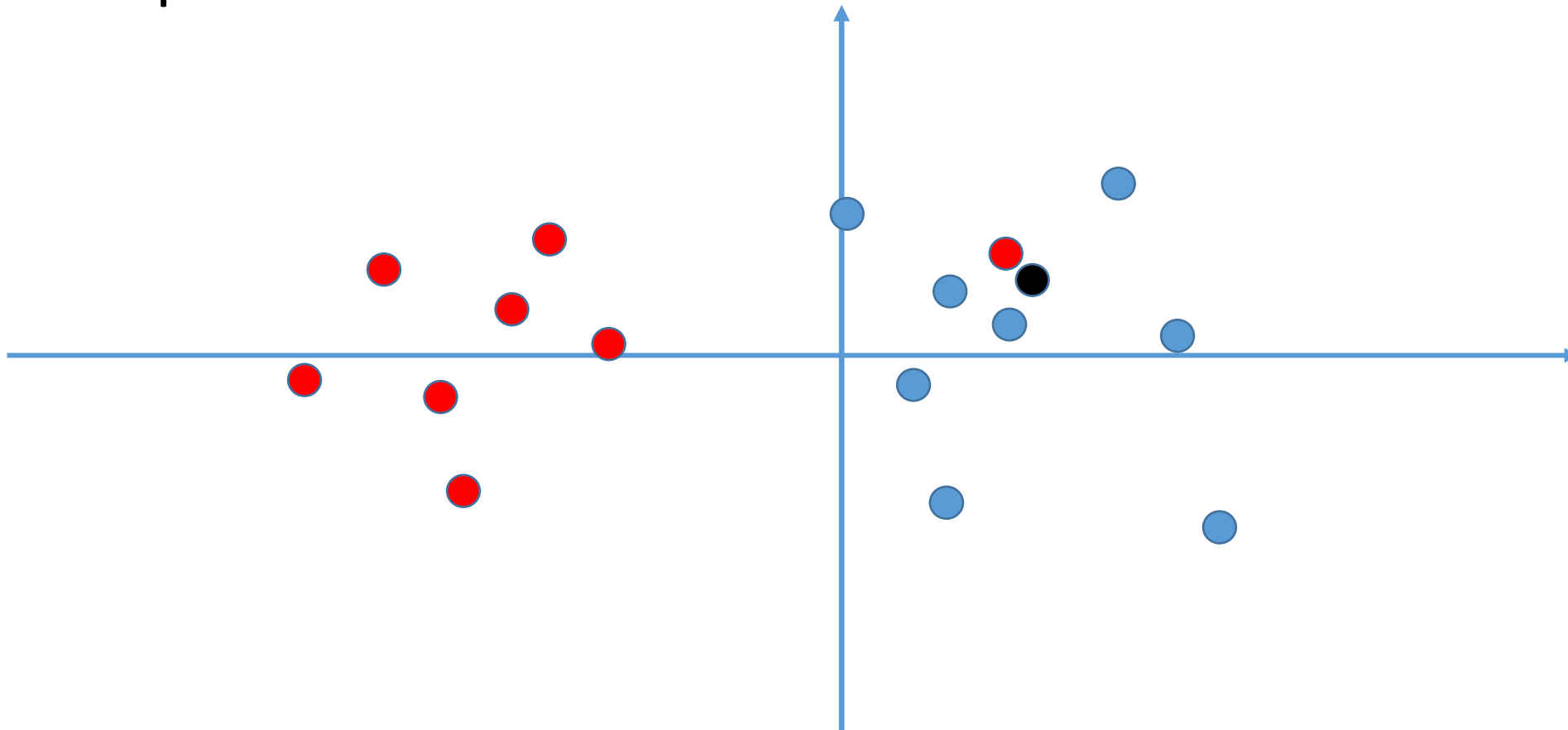
- Manhattan distance:

$$d(x_i, x_j) = \sum_f |x_{if} - x_{jf}|, \text{ where } x_{if} \text{ is the } f\text{-th feature}$$

How to pick the number of neighbors

- Split data into training and tuning sets
- Classify tuning set with different k
- Pick k that produces least tuning-set error

Example: the effect of k



What's the predicted label for the black dot using 1 neighbor? 3 neighbors?