

95-865: Support vector machines (another discriminative prediction method)

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Which decision boundary is best? SVM solution: maximize "margin" between red and blue points (make decision boundary line thicker until it hits a data point—this thickness is the size of the margin)

Decision boundary

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The points that the margin hits are called support vectors Which decision boundary is best? SVM solution: maximize "margin" between red and blue points (make decision boundary line thicker until it hits a data point—this thickness is the size of the margin)

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C-Support Vector Classification

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C-Support Vector Classification

- Basic version measures distance using Euclidean distance
 - Turns out to correspond to measuring similarity between two points by taking their dot product
 - This is called **linear svm**
- Can instead use a different similarity function ("kernel" function) instead (popular choice: Gaussian kernel, also called "radial basis function" kernel)
 - This is called **kernel svm**
- Also: support vector *regression* (these are all in sklearn)