

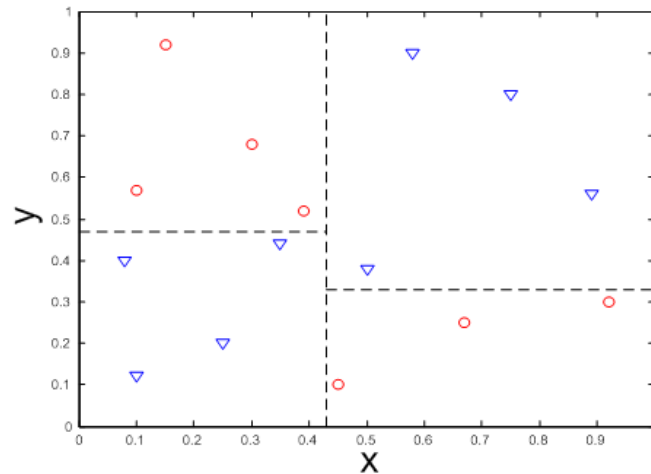
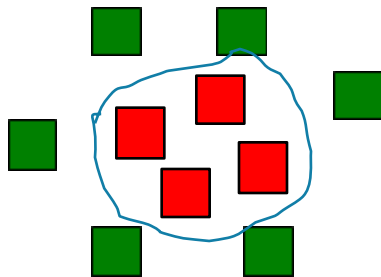


# MAX-MARGIN SVMs

**SYRACUSE UNIVERSITY**  
School of Information Studies

# THE SHAPE OF DECISION BOUNDARY

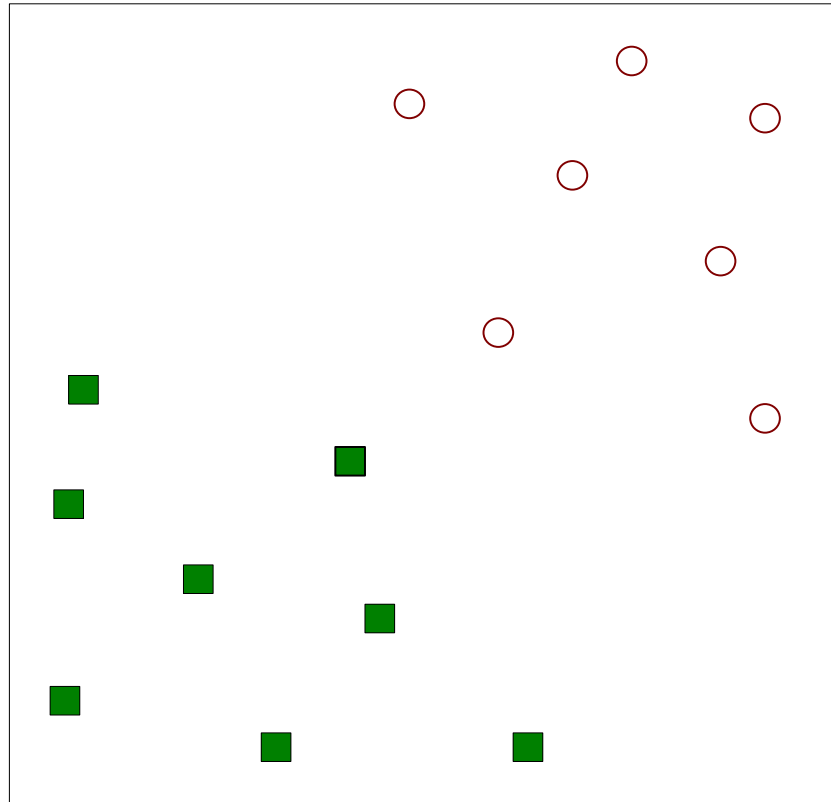
Some data are not linearly separable.



Support vector machine (SVM): An algorithm that can solve both linearly separable and inseparable problems

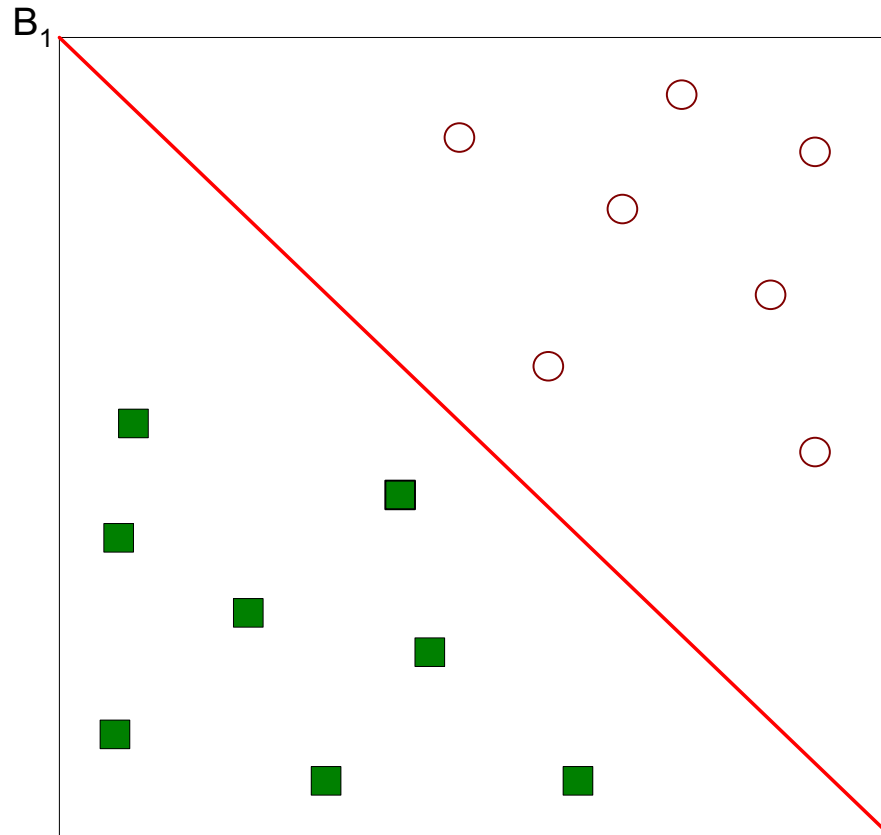
# SUPPORT VECTOR MACHINES

Find a linear hyperplane (decision boundary) that can separate the data.



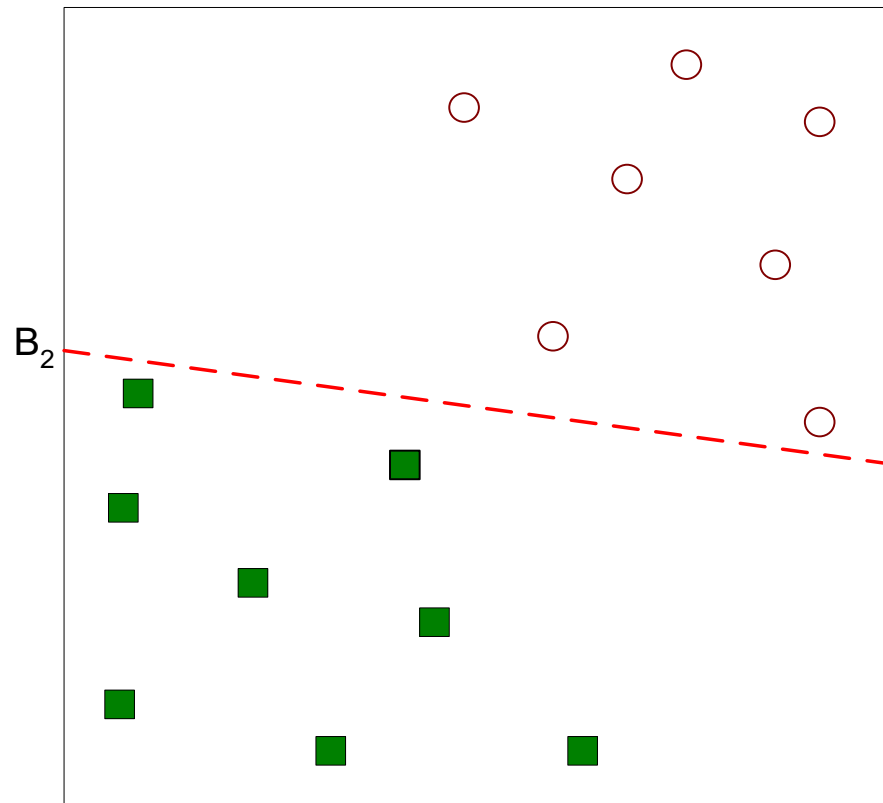
# SUPPORT VECTOR MACHINES

One possible solution:



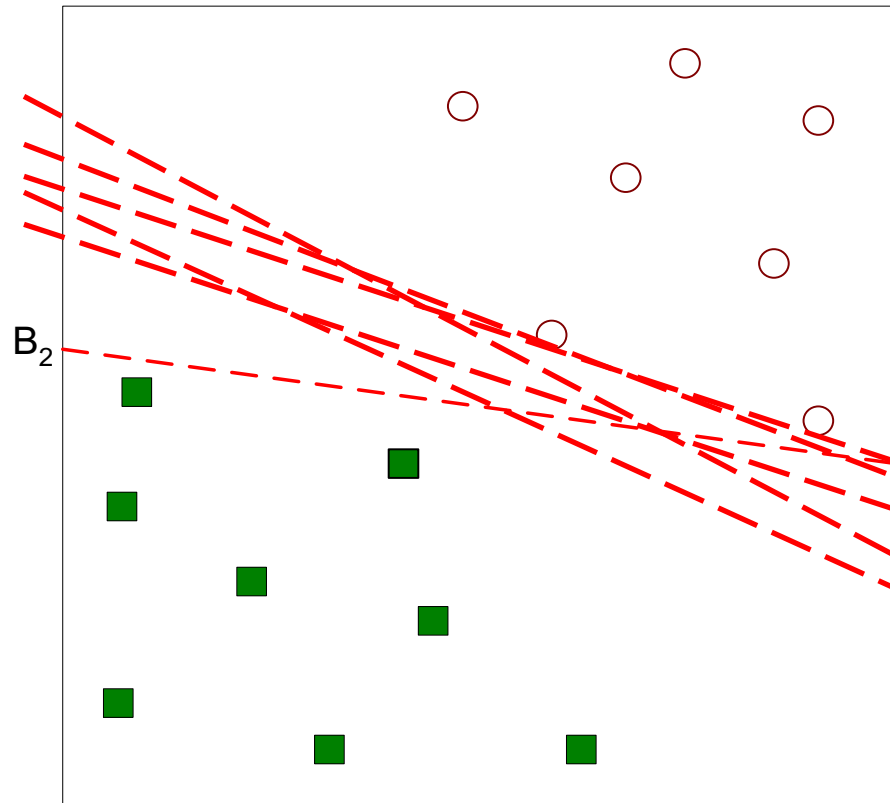
# SUPPORT VECTOR MACHINES

Another possible solution:



# SUPPORT VECTOR MACHINES

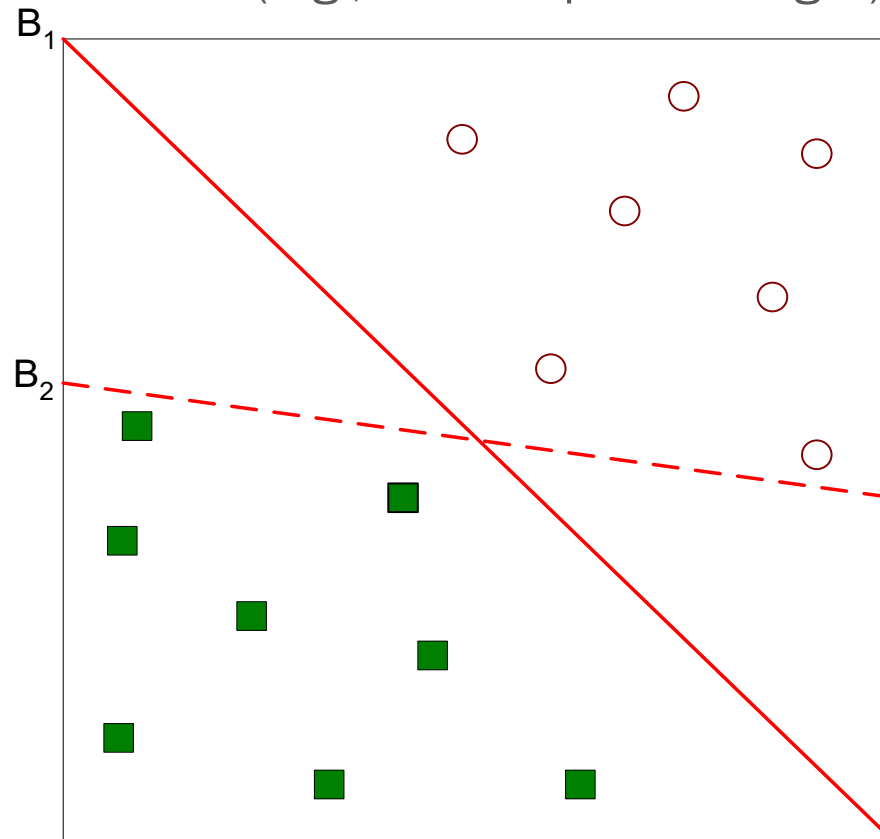
Numerous possible solutions:



# SUPPORT VECTOR MACHINES

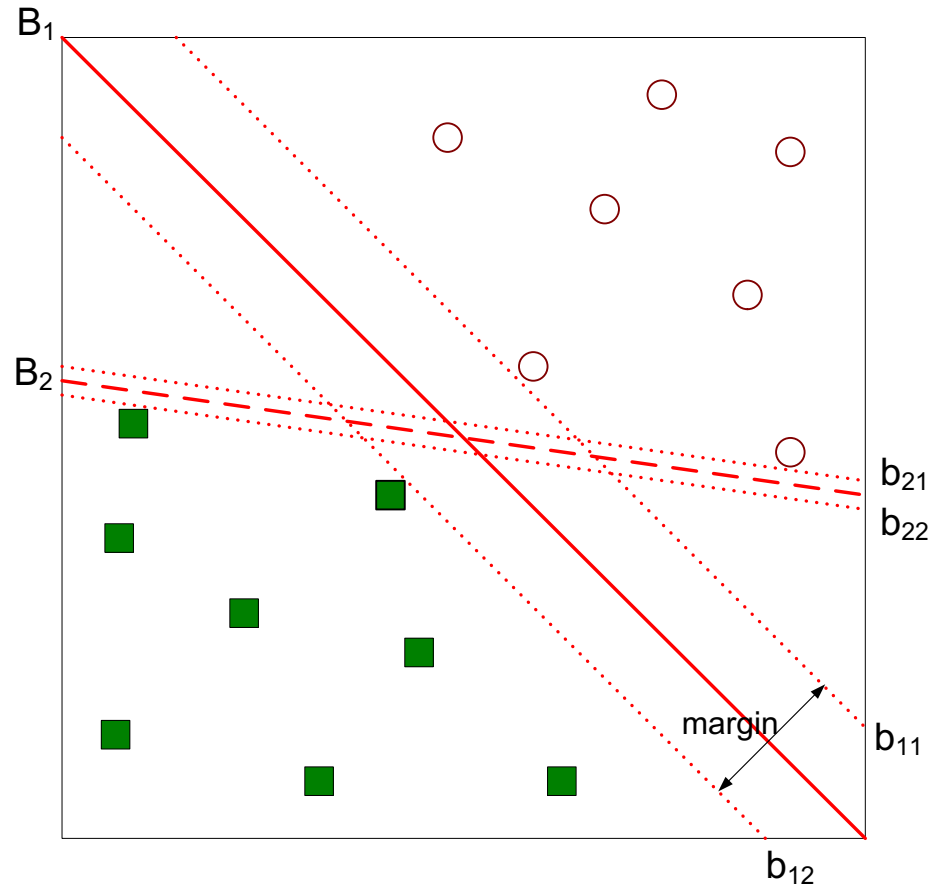
Which one is better: B1 or B2?

How do you define better? (E.g., “least square fitting”?)



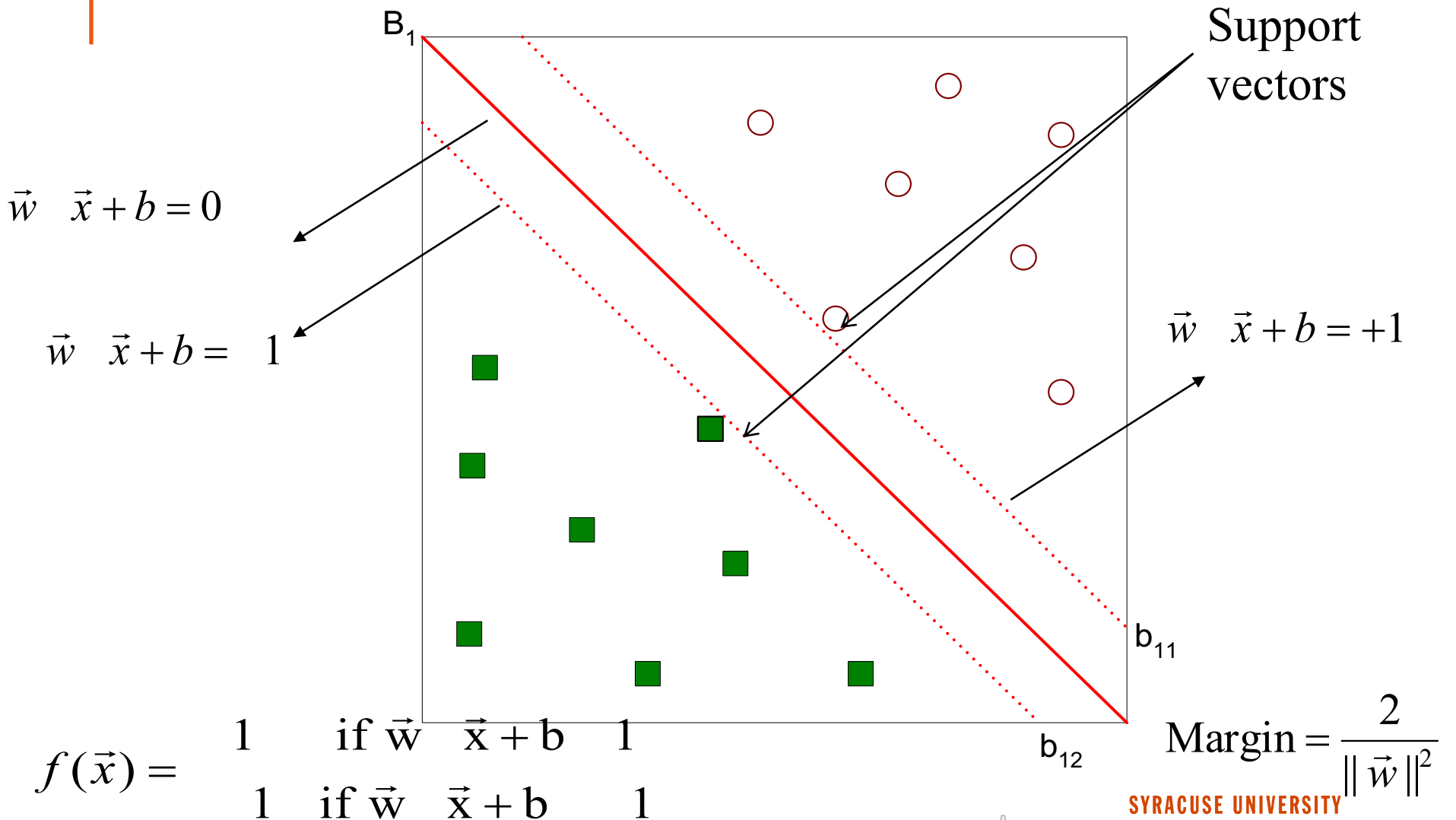
# SUPPORT VECTOR MACHINES

Find a hyperplane that **maximizes** the margin => B1 is better than B2.



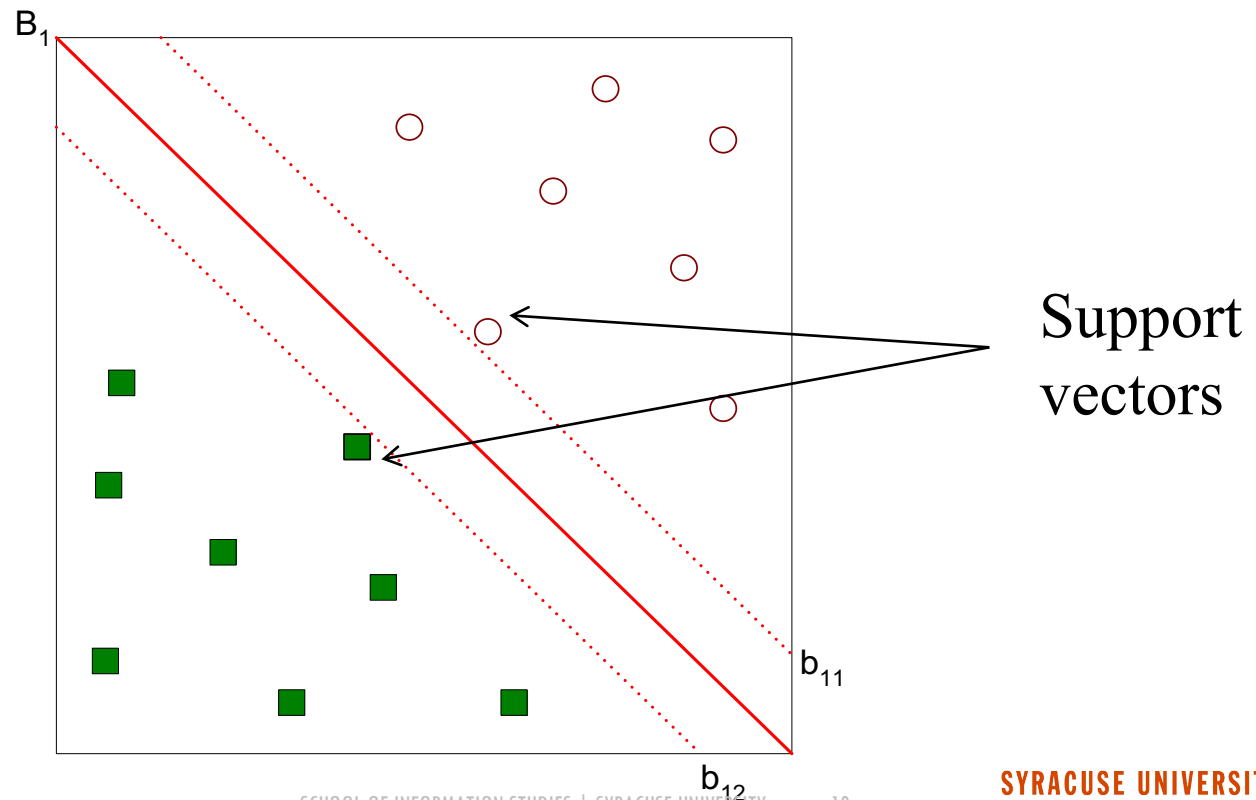


# SUPPORT VECTOR MACHINES



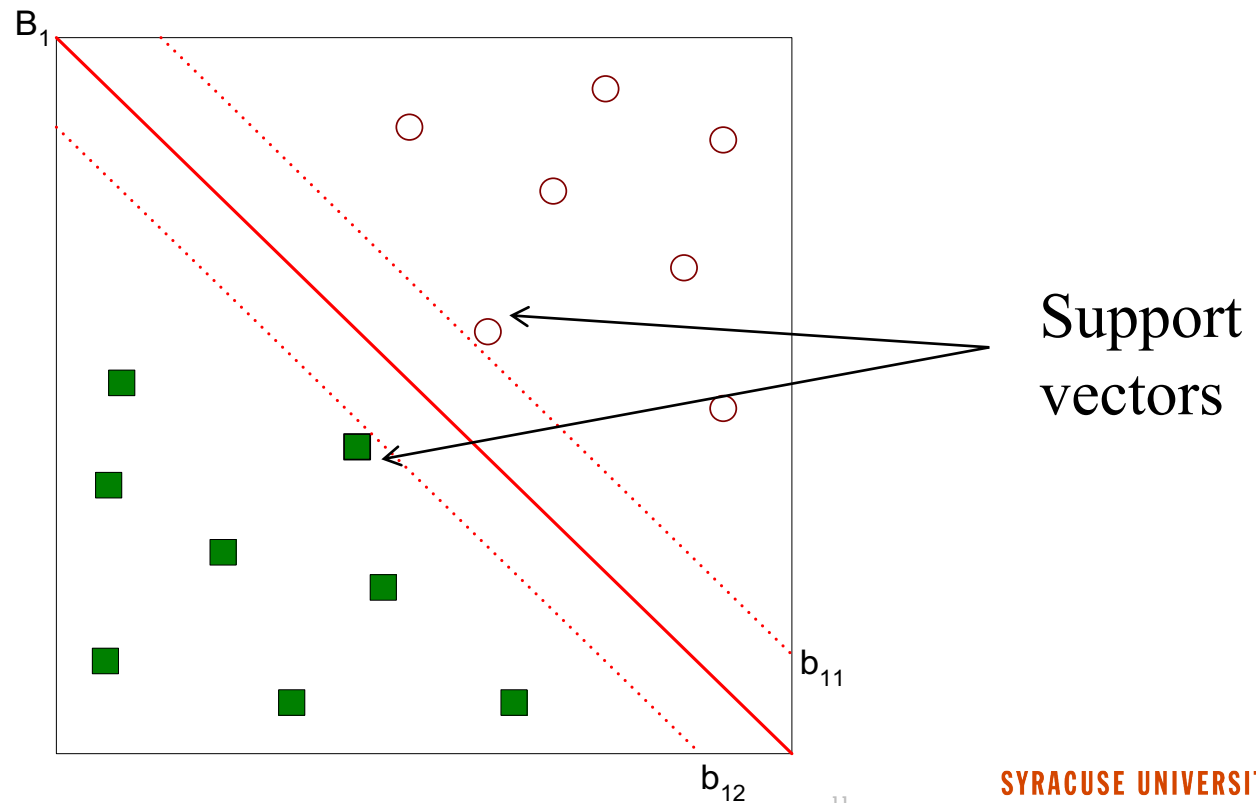
# SUPPORT VECTORS

Support vectors are the training examples (“vectors”) that are located on the margins.



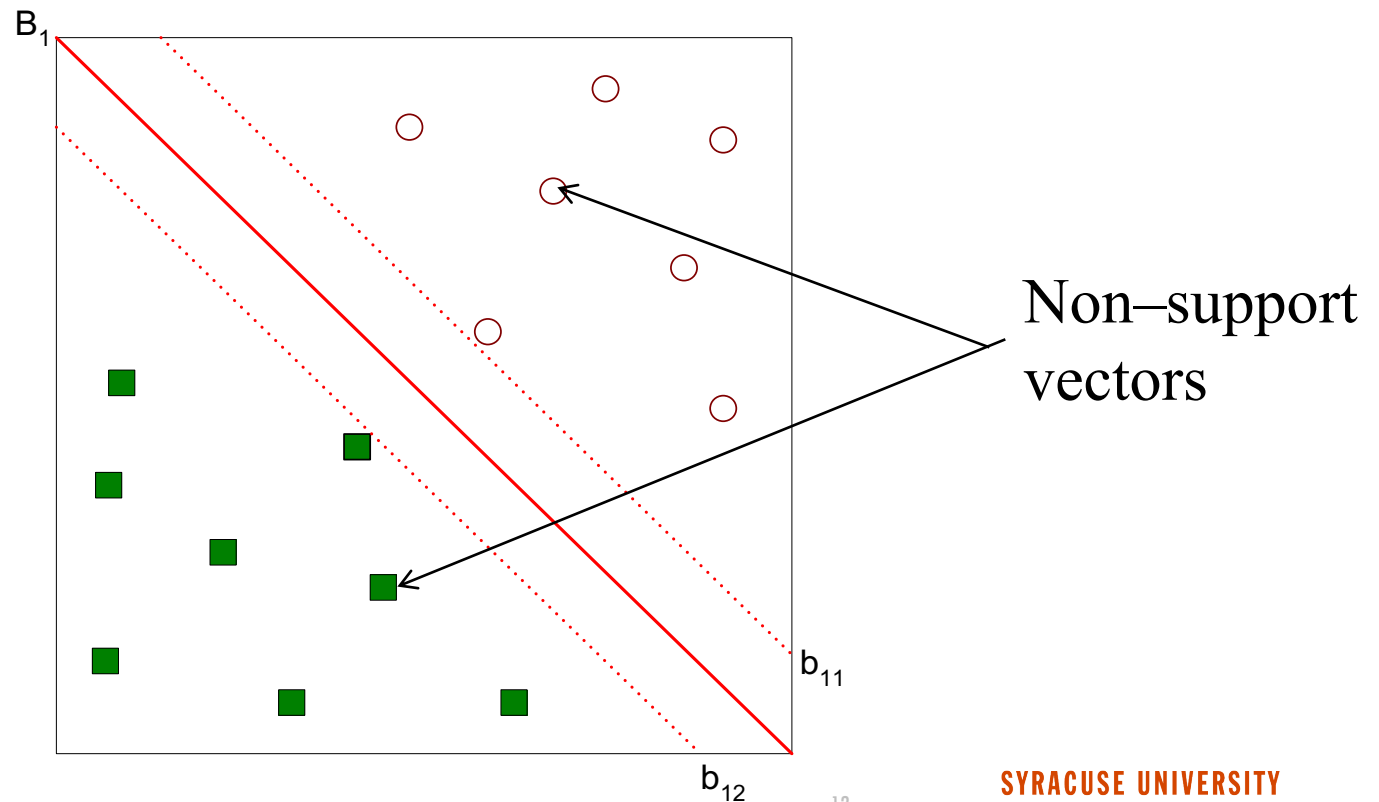
# DECISION BOUNDARY

Only support vectors determine the decision boundary.



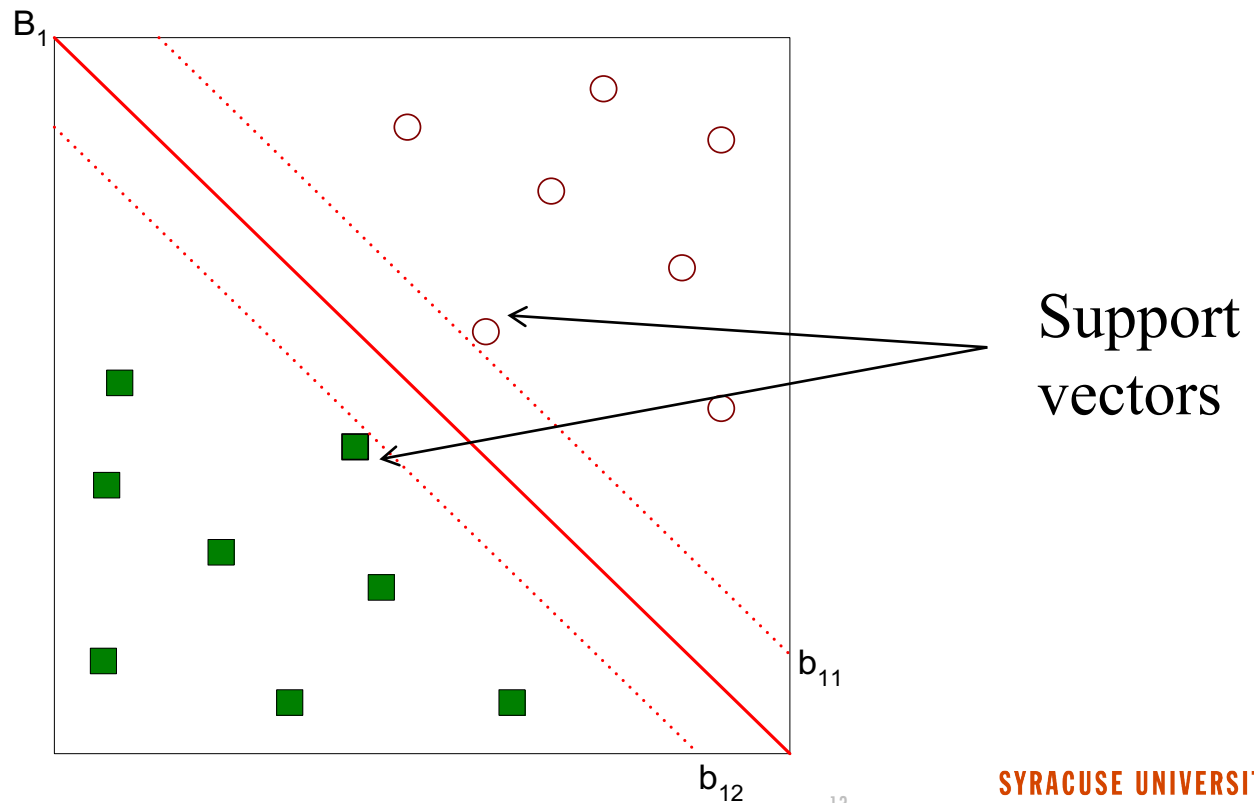
# NON-SUPPORT VECTORS

Training examples that are not support vectors do not participate in prediction.



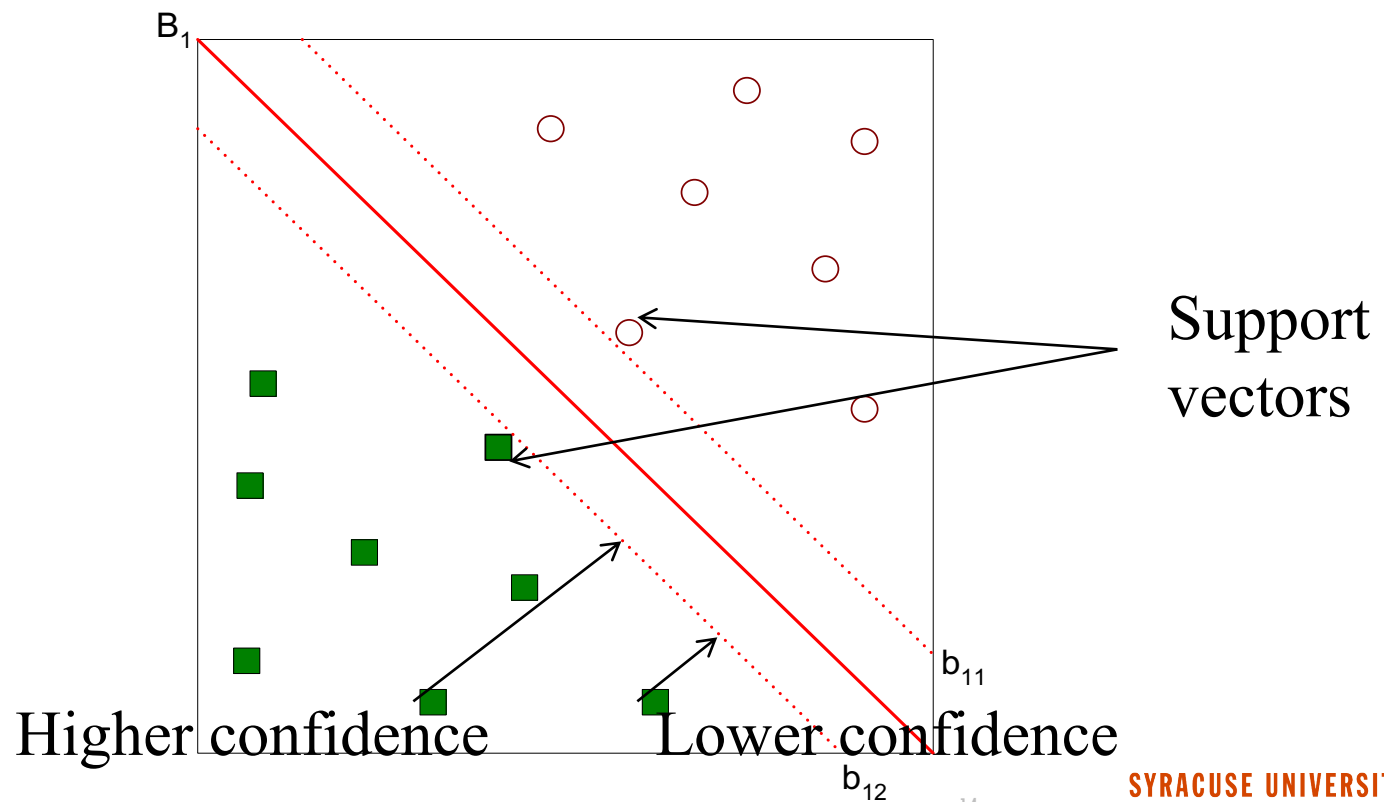
# MODEL COMPLEXITY

The number of support vectors is an indicator of the complexity of the trained SVM model.



# PREDICTION CONFIDENCE

The distance between the example and the decision boundary is an indicator of prediction confidence: The farther the better.



# PREDICTION CONFIDENCE

SVM's prediction result can be sorted by confidence and thus is suitable for semi-supervised learning and active learning.

Variant SVMs algorithm can be used for regression.