

If our current weight  $w_t$  makes a mistake on  $(x_i, y_i)$ , we perform perceptron update on  $w_t$  to get  $w_{t+1} = w_t + y_i x_i$ . What can we say about  $w_{t+1}$ ?

- A. It will now correctly classify  $(x_i, y_i)$
- B. It may correctly classify  $(x_i, y_i)$  but it is not guaranteed
- C. It now might misclassify another example that is previously correctly classified by  $w_t$
- D. while it does not guarantee correct classification of  $(x_i, y_i)$ , its prediction will move toward the correct direction, i.e., the value of  $w$  will increase.

Answers: B, C, D

# What does the convergence theorem provide us?

- A. The perceptron algorithm will always converge
- B. The convergence rate depend on the margin. It converges faster for data with large margin.
- C. The perceptron algorithm will converge to the max margin separator
- D. The perceptron algorithm convergence does not depend on the number of training examples or number of features

Answers: B, D