Machine Learning Lecture 3

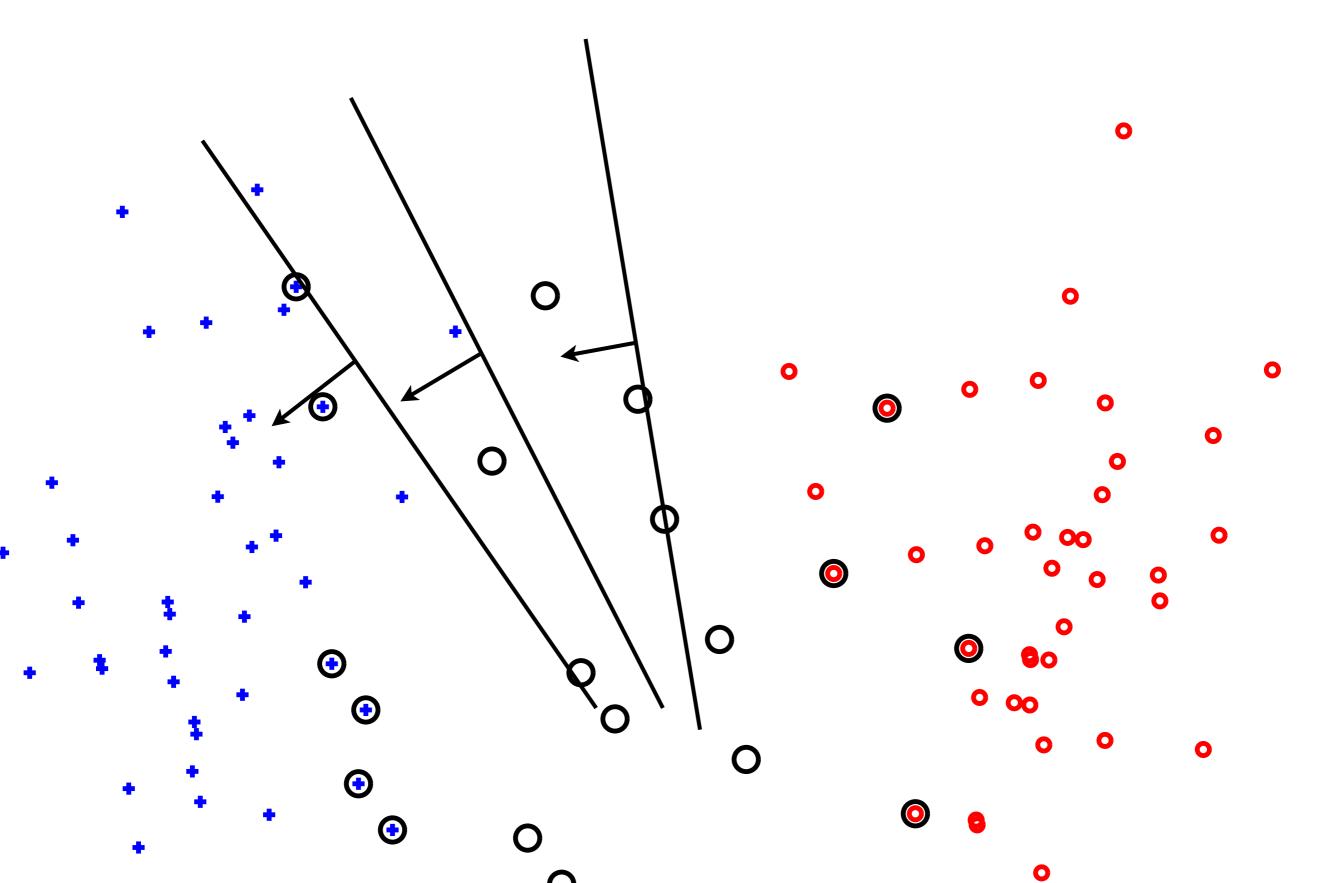


Outline

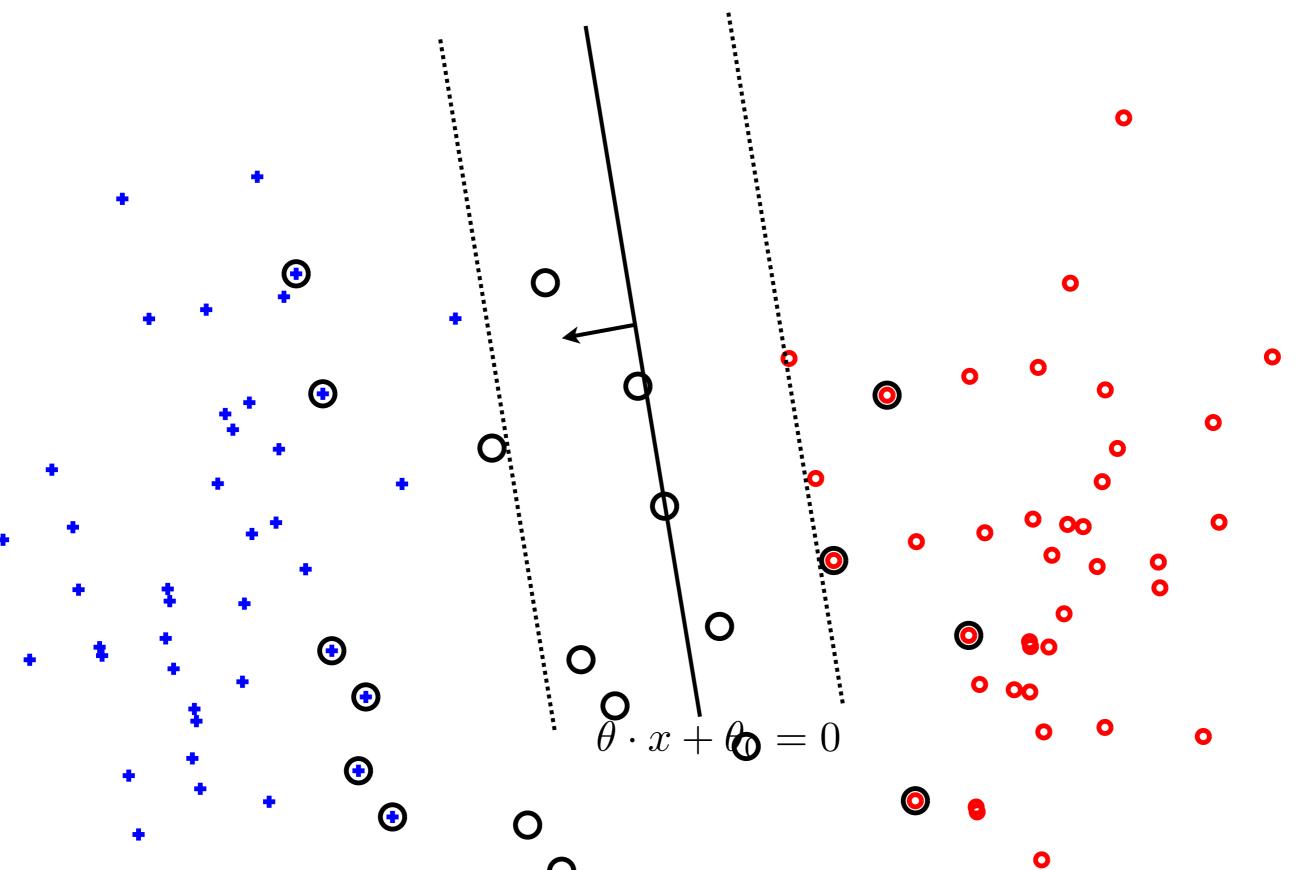
- Linear, large margin classification
 - margin, hinge loss, regularization
- Learning as an optimization problem



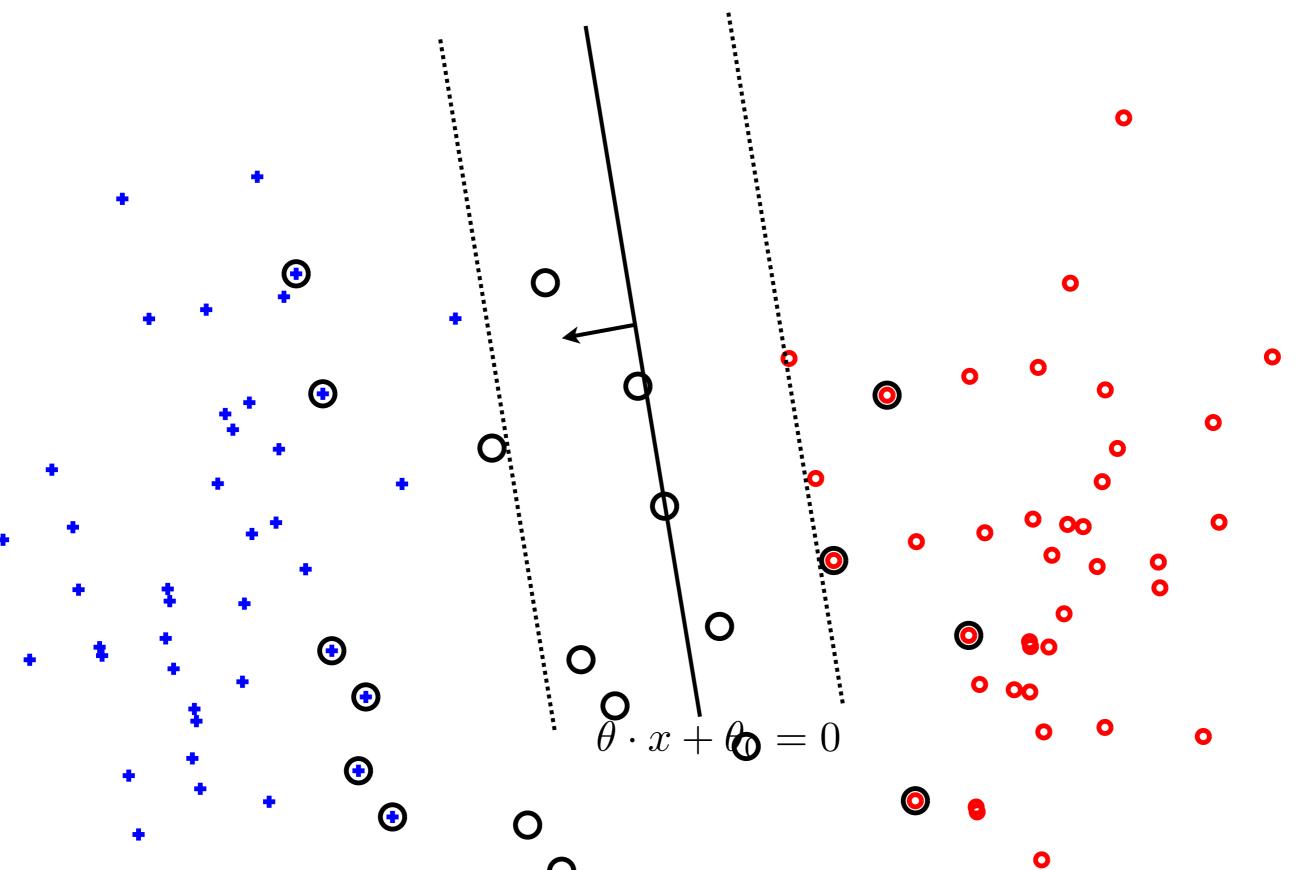
Linear classification



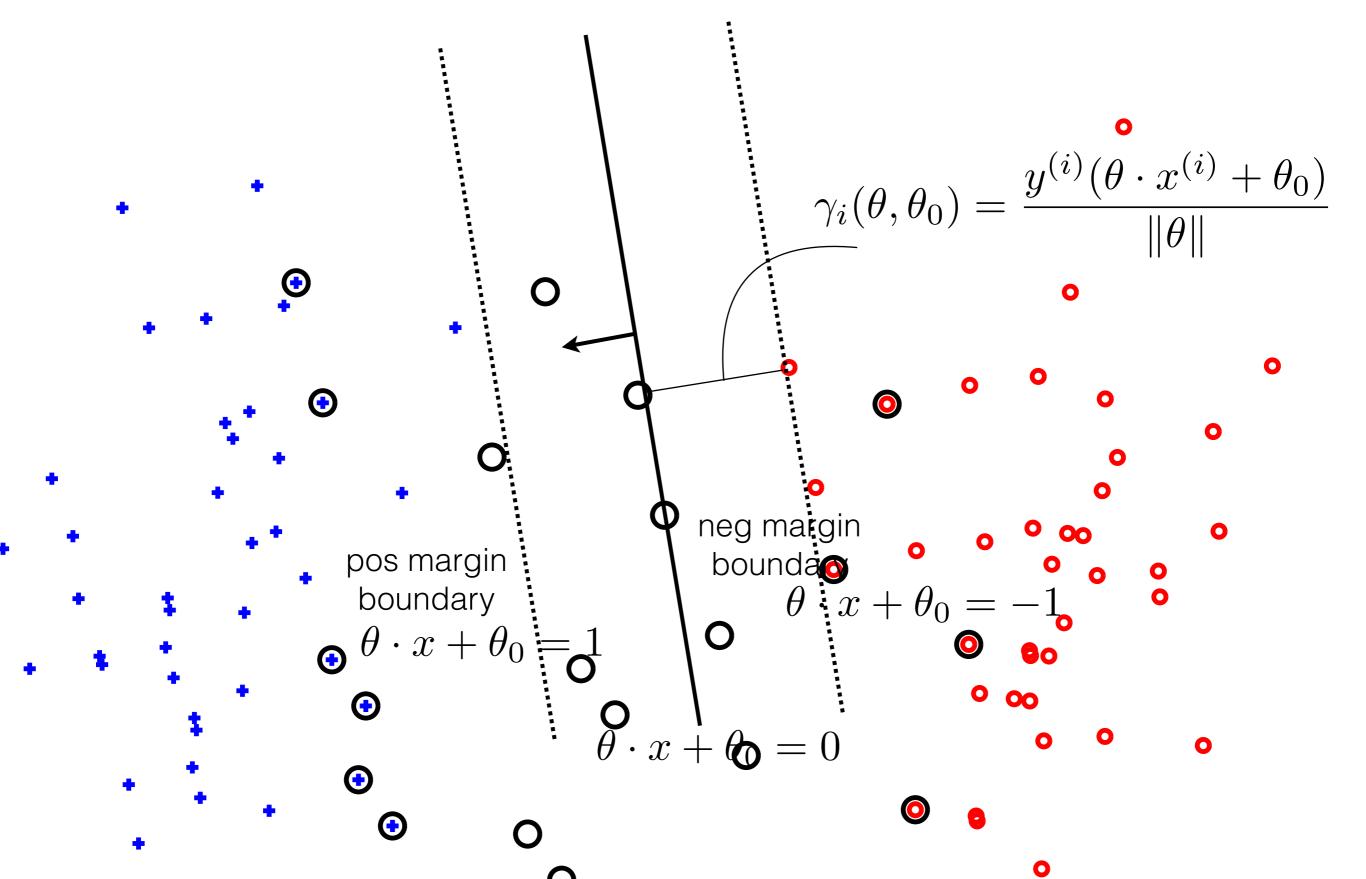








Linear classification, margin





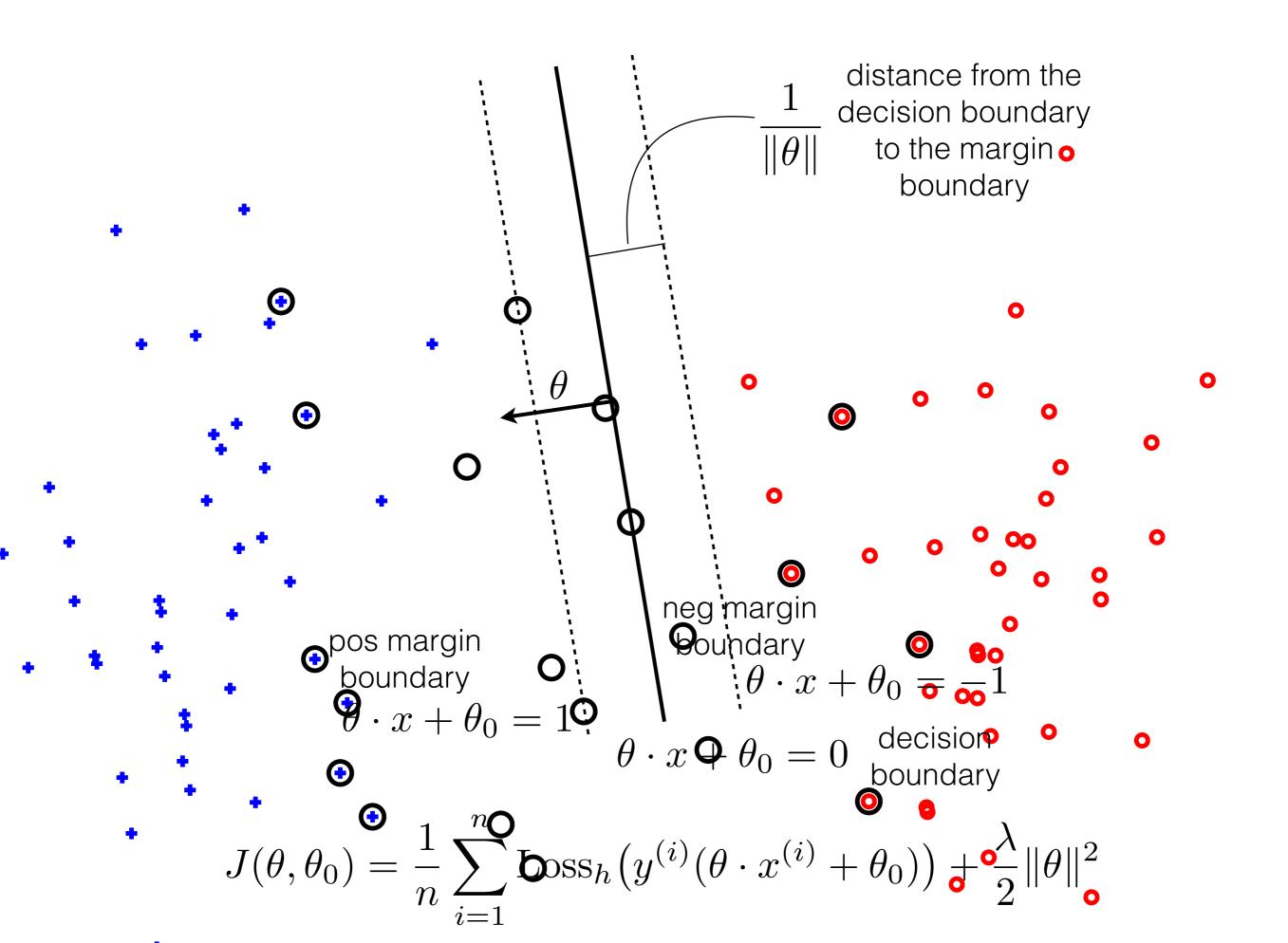
Hinge loss

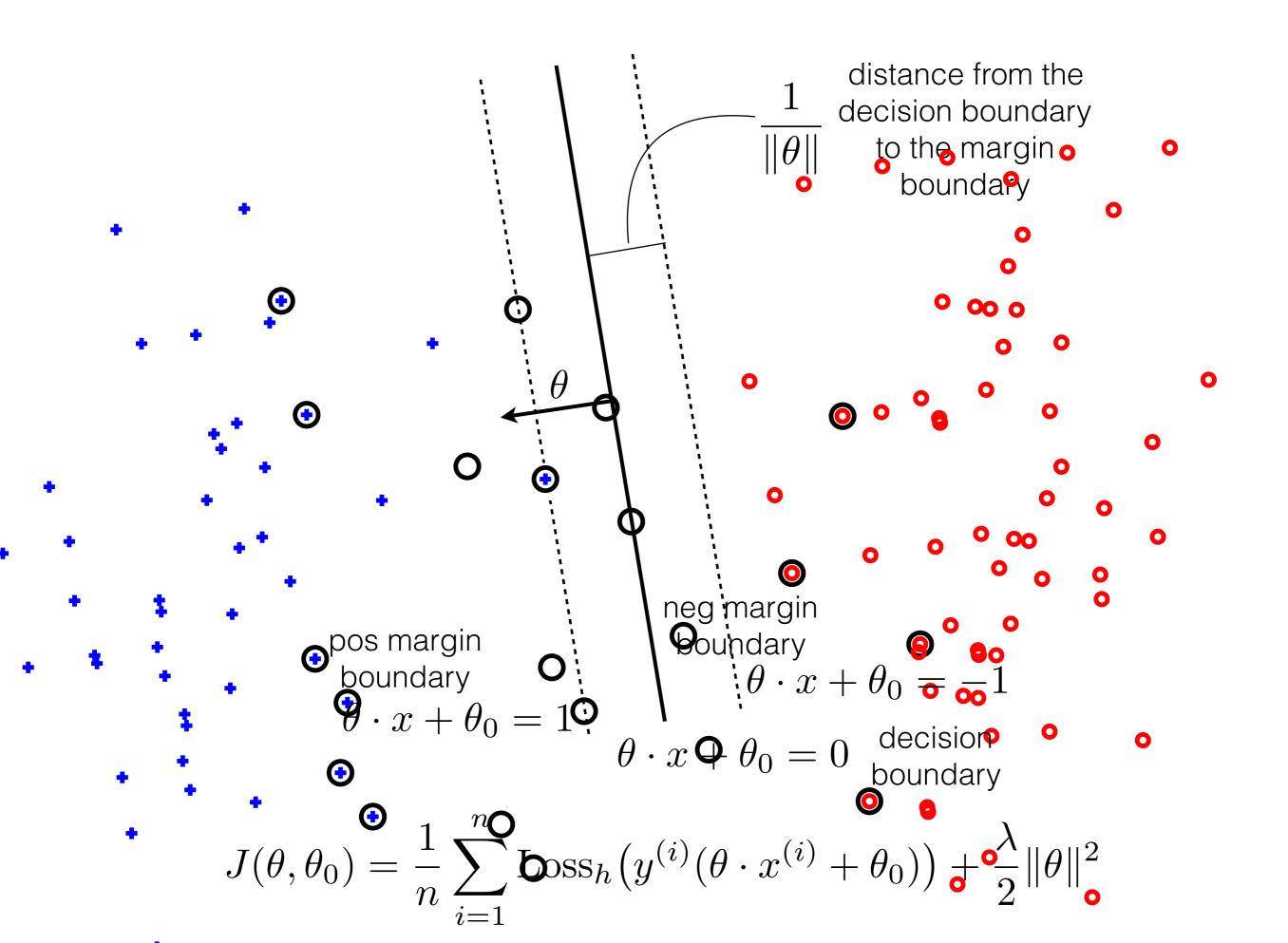
$$\operatorname{Loss}_h(y^{(i)}(\theta \cdot x^{(i)} + \theta_0)) =$$

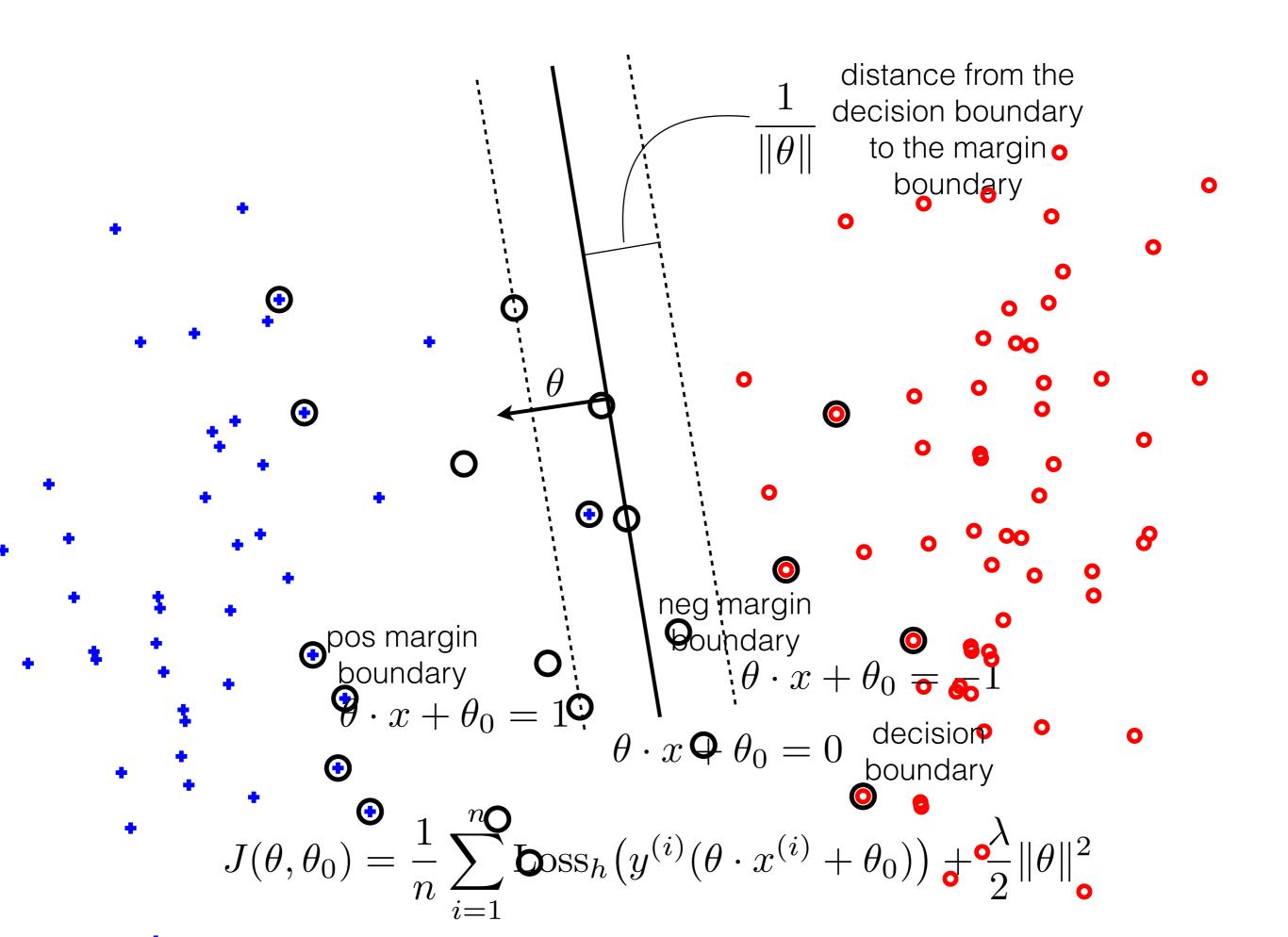
Regularization: towards max margin

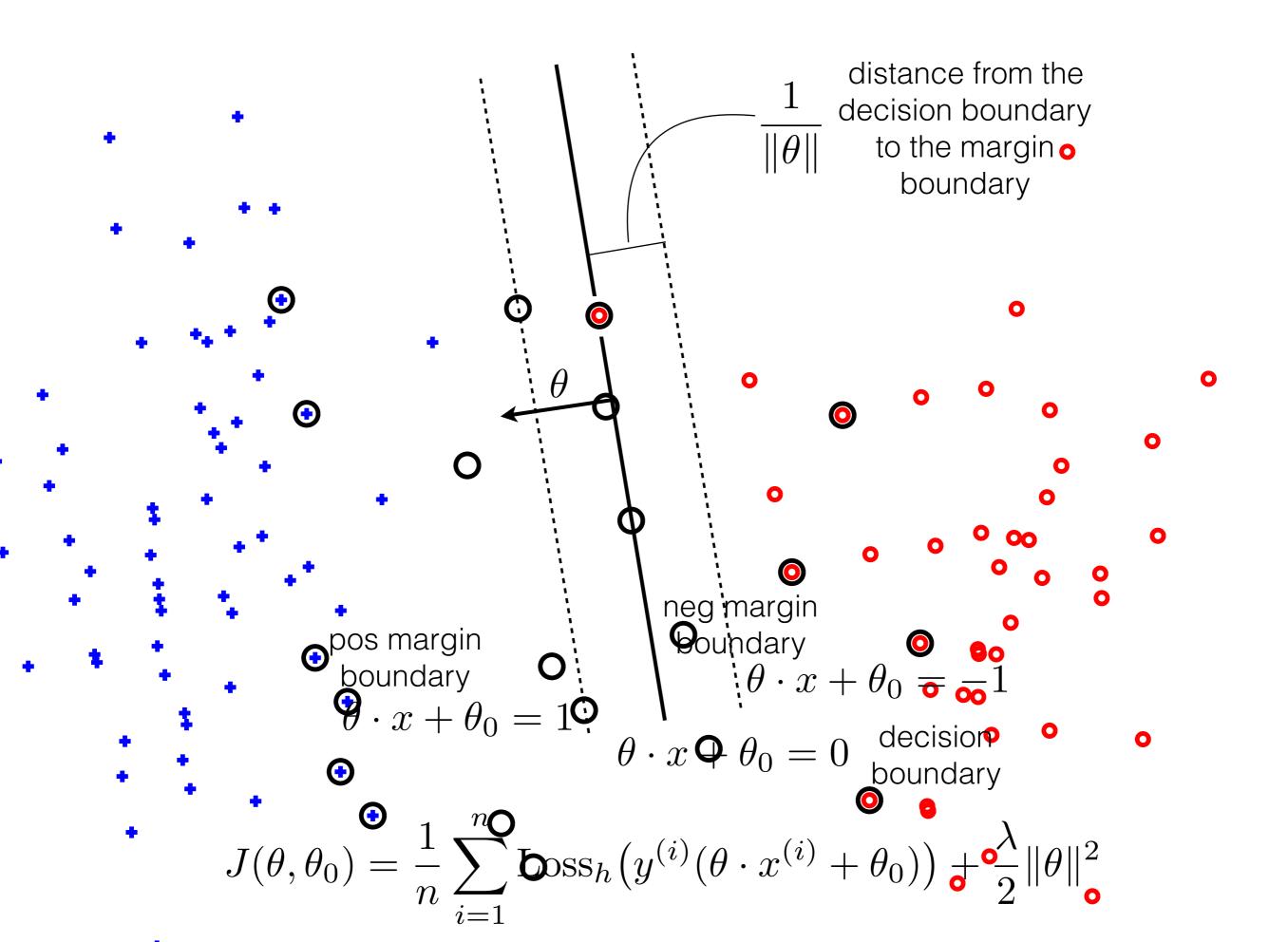
The objective

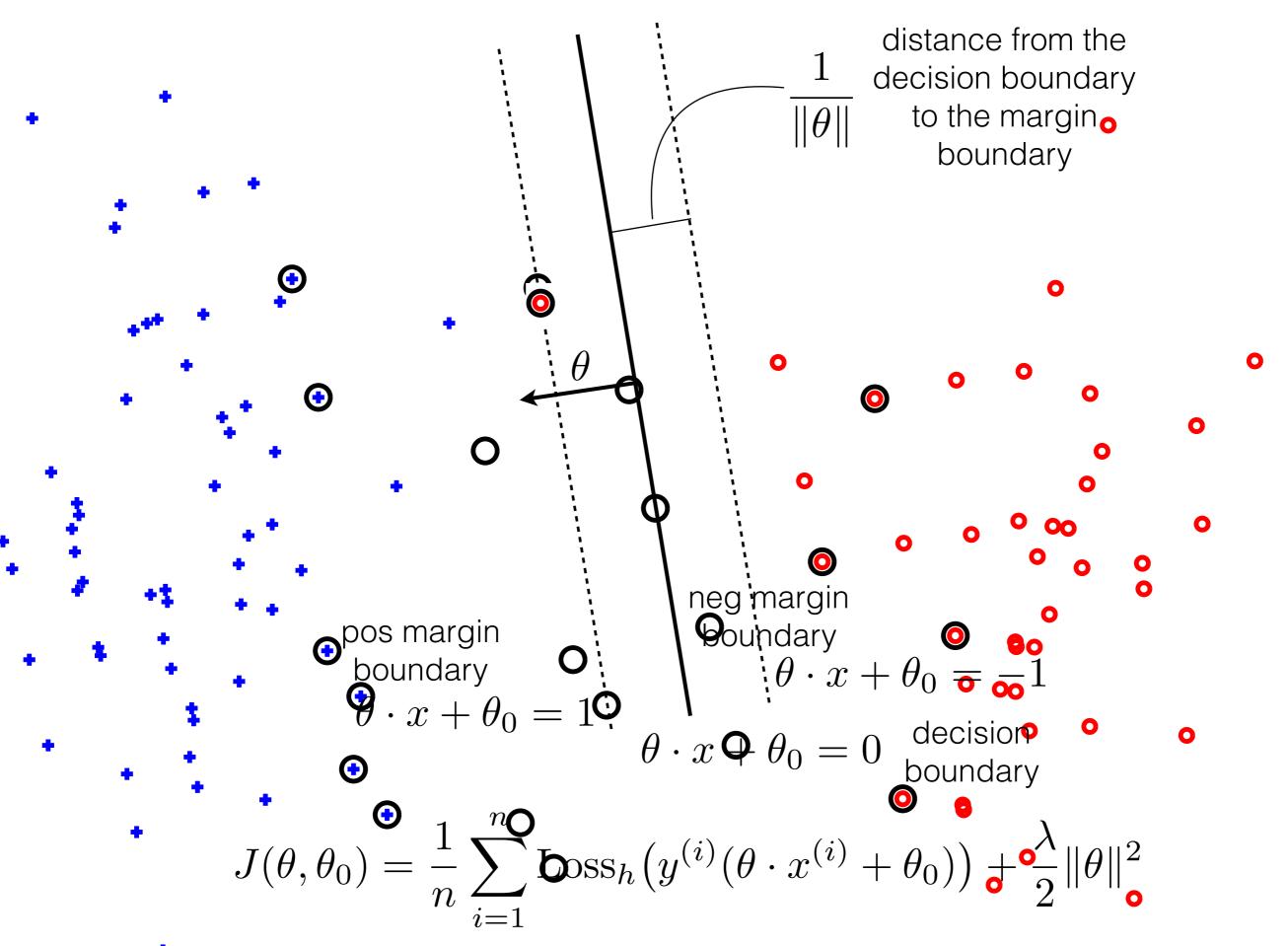
$$J(\theta, \theta_0) = \frac{1}{n} \sum_{i=1}^n \operatorname{Loss}_h \left(y^{(i)} (\theta \cdot x^{(i)} + \theta_0) \right) + \frac{\lambda}{2} \|\theta\|^2$$

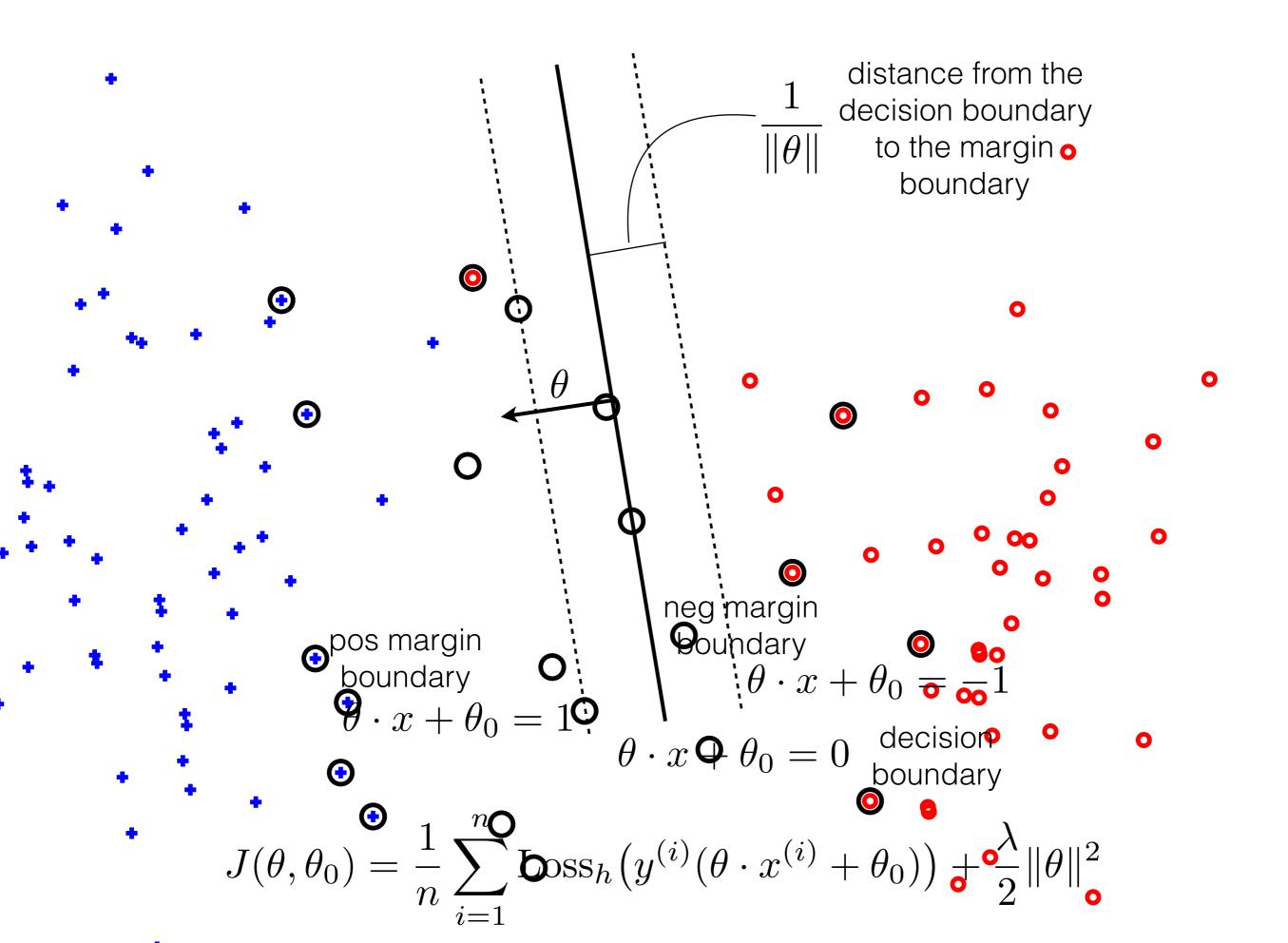


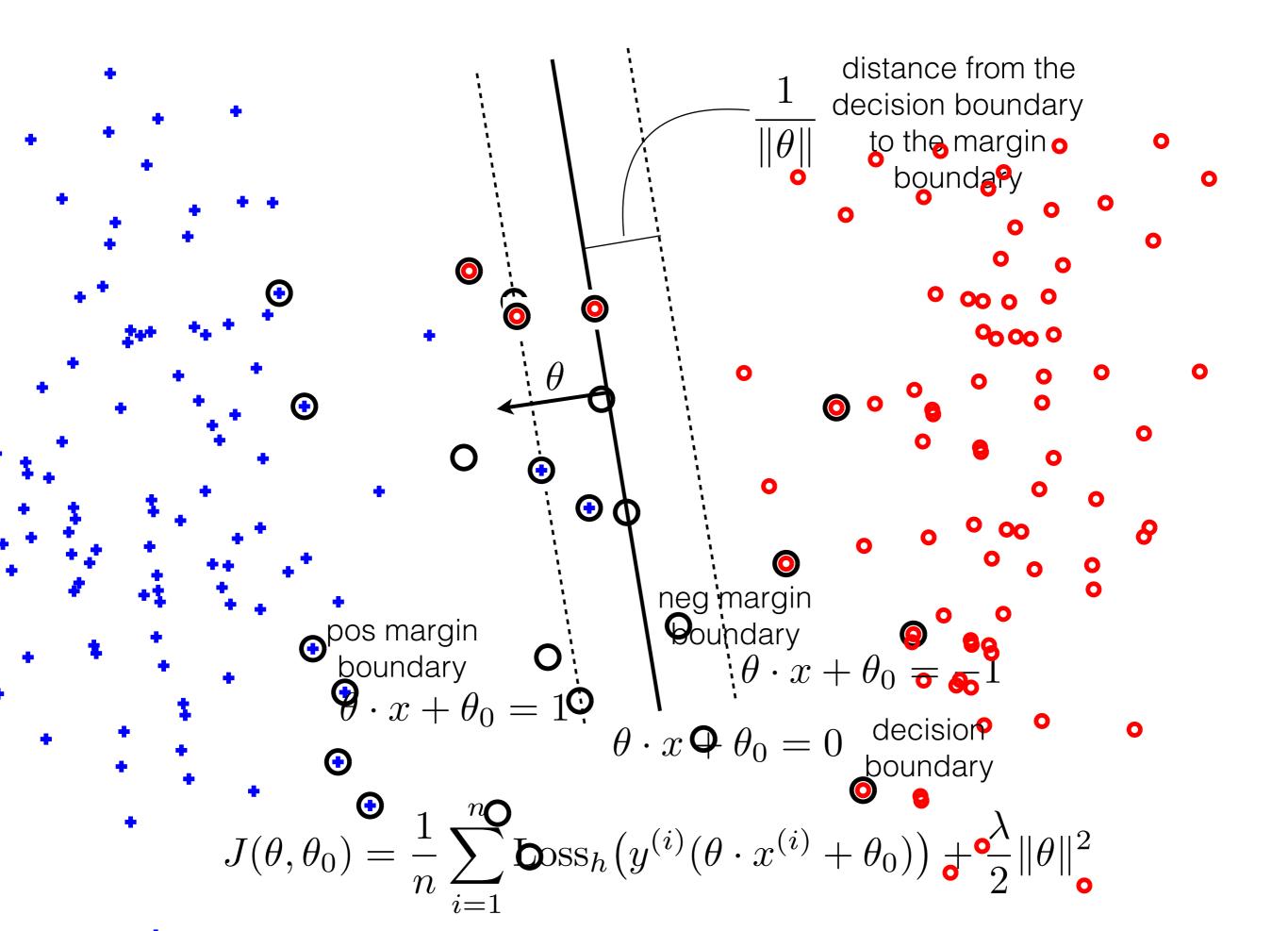














Things to know

General optimization formulation of learning

objective function = average loss + regularization

Large margin linear classification as optimization
margin boundaries, hinge loss, regularization

$$J(\theta, \theta_0) = \frac{1}{n} \sum_{i=1}^n \operatorname{Loss}_h \left(y^{(i)} (\theta \cdot x^{(i)} + \theta_0) \right) + \frac{\lambda}{2} \|\theta\|^2$$