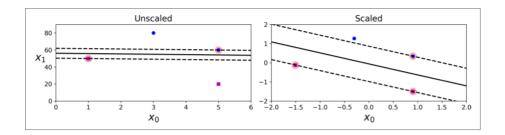
Support Vector Machines

Target: classification of complex but small- or medium-sized datasets

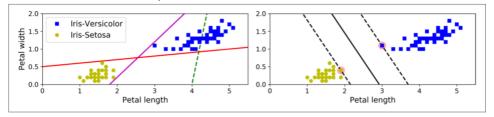
Goal: Find decision boundaries (hyperplane) for dataset

Notice: SVMs are sensitive to the feature scales (Scikit-Learn's StandardScaler)

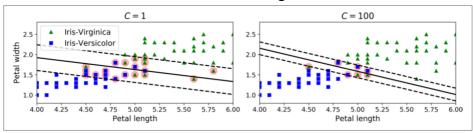


Types:

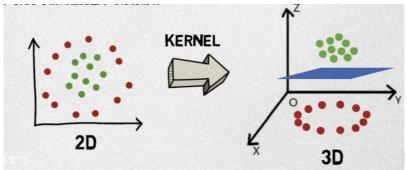
- Linear SVM Classification
 - Support vectors: decision boundary is fully determined by the instance located on the edge of the street (circled dot)
 - Large margin classification: find the widest possible street (dashed lines)
 between the classes, all instances should be off the street
 - Two issues:
 - data should be linearly separable
 - quite sensitive to outliers



- Soft margin classification: find a good balance between keeping the street as large as possible and limiting the margin violations(i.e., instances that end up in the middle of street or even on the wrong side)
 - C hyperparameter: control the balance, a smaller C value leads to a wider street but more margin violations



- Nonlinear SVM Classification
 - Kernel trick: project the instances in low dimensional space to those in high dimensional space



- Polynomial Kernel: as if you added many polynomial features, but without acutally having added them
 - Hyperparameter
 - degree: polynomial degree
 - coef0: contorls how much the model is influenced by highdegree polynomials versus low-degree polynomials
- o **Gaussian RBF Kernel**: add features computed using a similarity function that measure how much each instance resembles a particular landmark
 - Hyperparameter
 - Gamma (regularization term): a larger value makes each instances's range of influence smaller -> the decision boundary ends up more irregular, wiggling around individual instances

• SVM Regression

- essense: fit as many instances as possible on the street while limiting margin violation (i.e., instances off the street)
- o hyperparameter:
 - ε: control the width of the street
 - C: regularization term

