Supervised Learning: Linear and Non-Linear
optimal support Vector Machines X2 The support of the classes Molf-ple ways to separate the classes X1 Optimal hypoplane manually maximities
Optimal hyporplane maximites the wars of the separate the two groups. - Exactly et the middle of the two groups
Linear Support Vector Machines Popular approach in research, but not in industry Simplest case: maximize the margin the Distance between the boundary (hyperplane) and the support vectors (training examples closest to the boundary) Mak margin prefure not applicable in non-sperable case scilcit-learn: sklearn, sum. SVC
Some tows the classes can be separated, but not linearly: Non-Linear Support Vector Machines 1 Also popular in research, not industry "Kernelite" for non-linear problems - Choose a distance function named a Kernel" - Map the learning task to a higher dimension
Apply a linear classifier (SUM classifier) space in the new space Not newory-efficient, Secause it stores the support Jectors, which grow with the site of the training duta Computation 1s expensive
, sklaain, sum. SVC