Foundations of Machine Learning Al2000 and Al5000

FoML-15 Linear Discriminant Functions

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So far in FoML

- Intro to ML and Probability refresher
- MLE, MAP, and fully Bayesian treatment
- Linear Regression with basis functions (regularization, model selection
- Bias-Variance Decomposition (Bayesian Regression)
- Decision Theory three broad classification strategies
- Probabilistic Generative Models Continuous & discrete data





Discriminant Functions





Discriminant

- Function that takes an input and assigns one of the classes as output
- Restrict to 'Linear Discriminants'
 - Decision surfaces are hyperplanes





- Input $x \in \mathbb{R}^D$
- Targets $t \in \{C_1, C_2\}$
- Discriminant Function

$$y(\mathbf{x}) = f(\tilde{\mathbf{w}}^T \phi(\mathbf{x}))$$

 $\phi(\mathbf{x}) = (\phi_0(\mathbf{x}), \dots, \phi_{M-1}(\mathbf{x}))^T$

Linear Decision boundaries

Generalized Linear Models

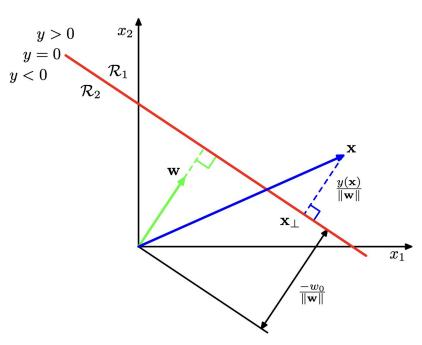




- Simplest discriminant function $y(\mathbf{x}, \tilde{\mathbf{w}}) = \mathbf{w}^T \mathbf{x} + w_0$
- Decision boundary $y(\mathbf{x}, \tilde{\mathbf{w}}) = 0$



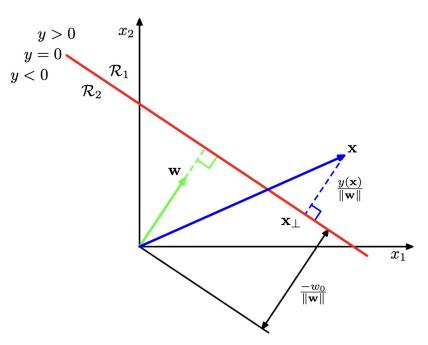




W determines the orientation of the decision boundary





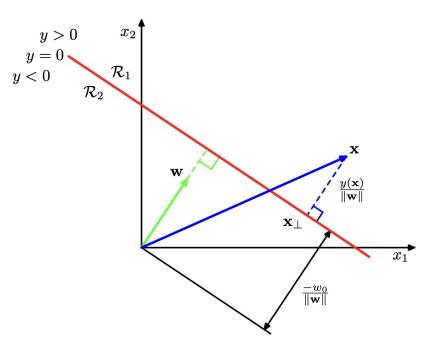


Normal distance from origin

W₀ shifts the boundary away from origin



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y(x) gives the signed perpendicular distance from the boundary

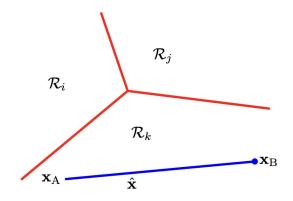




Discriminant Functions: Multiple Classes

- K-class discriminant $y_k(\mathbf{x}) = \mathbf{w}_k^T \mathbf{x} + w_{k0}$
- Class assignment
 - o to C_ν if
- Decision boundary:

Decision regions (for GLM) are convex







Next Least Squares for Classification



