

# Linear Methods for Regression

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October 20, 2006



# Statistical Decision Theory: Versions of Expected Losses

[Point Estimation Problem]

Let  $X_1, \dots, X_n \sim f_\theta$ , for example, pdf of  $N(\theta, \sigma^2)$  or pmf of  $Bernoulli(\theta)$  ObjectIFind  $\hat{\theta}_*$







# Complete Class Theorem

- ▶ A class  $\mathcal{C}$  is *complete* if for any decision  $\delta$  not in  $\mathcal{C}$ , there exists a decision  $\delta'$  which dominates  $\delta$ .
- ▶ Under some regularity conditions, the class of generalized Bayes procedures form a complete class.
- ▶ Implication: Search no further. Work with generalized Bayes procedures.

$(Y|X)$

Consider  $X \in R^p$ ,  $Y \in R$  with joint prob distribution  $P_{Y,X}$ . Seek a ftn  $f$  for predicting  $Y$  given  $X$ . Under  $L(Y, f(X)) = (Y - f(X))^2$ , squared error loss, in the spirit of Bayes risk, find  $f$  min

$$EPE(f) = E(Y - f(X))^2 = \int (y - f(x))^2 dP_{Y,X}(y, x) \quad (1)$$



## KNN as $(Y|X)$

Let  $T = \{(Y_i, X_i)\}_i^n$  be the training data.

- ▶  $\hat{f}(x) = \text{Ave}(y_i | x_i \in N_k(x))$ , where “Ave” = average,  $N_k(x)$  is the neighborhood containing the  $k$  points in  $T$  closest to  $x$ .
- ▶ expectation is approximated by averaging over sample space.
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- ▶ Regression is approximated by averaging over sample space.
- ▶ Conditioning at a point is like conditioning on some region “close” to the target point.

How good is NN? Rationale: Series is over

- ▶ Sample size usually small
- ▶ As  $p$  increases,  $N_k(x)$  becomes huge
- ▶ Convergence
  - ▶ Convergence holds.  $\hat{f} \rightarrow f$  as  $n \rightarrow \infty$
  - ▶ Slower rate of convergence.

# LS as $(Y|X)$

- ▶  $f(x) \approx x^T \beta$
- ▶ Plug this  $f$  into

# KNN vs. LS

- ▶ LS assumes  $f(x) \approx$  globally by a linear function
- ▶ NN assumes  $f(x) \approx$  locally by a const function
- ▶ Functional approximation

