

# PA 2018 -02

---

## Remark : Least squares decision boundary

---

KNN and DB

3-NN vs 1-NN

Bias-Variance tradeoff

## PA Mind Map

---

Pattern analysis

- Density Estimation
  - Task: To estimate a continuous Probability density function from a discrete finite set of observations
  - eg:
    - MLE(Maximum Likelihood Estimation):
      - Assume that PDF is a Gaussian,Poisson,gamma,polynomial distribution
      - Estimate mean and covariance
      - More General: Assume any parametric function
    - MAP(Maximum a posteriori Estimation):
      - similar to MLE, but with prior knowledge
      - $\theta^* = \operatorname{argmax}(\theta | x_1, x_2, x_3 \dots x_n)$ , where  $\theta$  denotes the parameter of the model (eg mean and covariance for a Gaussian), and  $x_1 \dots x_n$  denotes N observations/samples/feature points
      - $\operatorname{argmax}_{\theta} \frac{p(\theta) * p(x_1, x_2, x_3 \dots x_n | \theta)}{p(x_1, x_2, x_3 \dots x_n)}$
      - assume independence of observations!
      - $\operatorname{argmax}_{\theta} p(\theta) * \prod_{i=1}^N p(x_i | \theta)$
      - $\operatorname{argmax}_{\theta} \log p(\theta) + \sum_{i=1}^N \log(p(x_i | \theta))$
- Mean Shift algorithm
- Clustering
  - Group observations according to vicinity in the feature space
- Manifold Learning
  - Dimensionality reduction
  - PCA(Principal component Analysis): Linear projection of the d-dimensional data onto a  $d_{-}$ -dimensional subspace that maximizes the covariance of the data.

up now is "Roughly analytical"

- Random Forests
  - Learning-based probabilistic tool to partition the feature space
- Graphical Models HMMs,MRFs
  - To model the sequence of data,
  - Hidden Markov Models

- Markov Random Fields