

## Lecture 15

### Bayesian Reasoning

### Linear Least Squares

### State Space Control Principles

- Controllable
- Observable
- Luenberger observer

### Regression

### Expected Value, Variance

- Matrix version
- Scalar version

### Fusion (optimal linear mix) + Scalar Kalman

### How to Practice

### Bayesian Reasoning Example

- $P(A | B) \rightarrow B$  has happened
- $P(A|B) = \frac{P(A \cap B)}{P(B)}$
- $P(A|B) = \frac{P(B|A)P(A)}{P(B)}$
- Partitioning is when you split into sections
  - $\cup B_k = u$
  - $B_k \cap B_l = \text{Disjoint}$
  - $P(k) = \sum P(A|B_k)P(K_k)$

### Linear Least Squares Example

- $(x^k, y^k)$
- Linear Model  $F(x) = \sum \alpha_k J_k(x)$
- $\frac{\text{Minimize}}{\alpha_1 \alpha_n} \sum_{k=1}^M \|F(x^k) - y^k\|$
- Given  $(x^k, y^k)$ ,  $k=1 \dots N$
- Model is  $f(x) = aX$
- $\begin{bmatrix} x^1 \\ x^2 \\ x^N \end{bmatrix} * [a] = \begin{bmatrix} y^1 \\ y^2 \\ y^N \end{bmatrix}$
- $A^T A x = A^T y$
- $\begin{bmatrix} x^1 & \dots & x^n \end{bmatrix} * \begin{bmatrix} x^1 \\ \dots \\ x^N \end{bmatrix} a = \begin{bmatrix} x^1 & \dots & x^n \end{bmatrix} y$
- $\sum x_k^2 a = \sum x_k y_k$
- $a = \frac{\sum x_k^2}{\sum x_k y_k}$

### Probability Example

- Random Var  $X \rightarrow x = \text{scalar}$
- Random Vector  $(X_1, X_2) = \text{matrix}$
- $E(x) \rightarrow \text{Definition } u(x) = \frac{1}{N} \sum x^k$
- $VAR(x) = E((E(x) - x)^2)$
- $CO - VAR(x, y) = E((E(x) - x)(E(y) - y))$

### Scalar Example

- $x$  is a random variable,  $C$  is a constant

- $\text{VAR}(x + C) = \text{VAR}(x)$
- $f(x)$  is a linear mapping
- $\Re - > \Re$
- I know the variance of  $x$ .
- What is the  $\text{VAR}(f(x))$
- $f(x) = ax$
- $\text{VAR}(ax) = a^2 \text{VAR}(x)$

#### Vector Example

- $\text{VAR}(X) = E((E(x) - x)(E(x) - x)^T)$
- $(AB)^T = B^T A^T$
- $X^T y = y^T X$

#### State Space Control Example

- Controllable, Observable
- $x_{m+1} = Ax_m + Bu_m$
- $y_m = Cx_m$