

EE16A - Lecture 22 Notes

Name: Felix Su SID: 25794773

Spring 2016 GSI: Ena Hariyoshi

Least Squares Geometric Argument

1. Best estimate (1D): $\hat{b} = \frac{\vec{a}^T \vec{b}}{\vec{a}^T \vec{a}} \vec{a}$ where $\frac{\vec{a}^T \vec{b}}{\vec{a}^T \vec{a}}$ = best scalar x
 2. Best estimate (Multi-D): $x = (A^T A)^{-1} A^T b$
 3. Error Vector: $\varepsilon = \vec{b} - A\vec{x}$ because $\vec{b} \notin \text{span}\{\vec{a}_1, \dots, \vec{a}_n\}$
 4. Goal: Minimize $\|\varepsilon\|^2$ over all possible \vec{x}
 5. Best \vec{x} ensures that $\varepsilon \perp \sum_{i=1}^n \alpha_i \vec{a}_i \forall \alpha_i \in \mathbb{R} \implies \varepsilon \perp a_n$ where a_n = each column vector of A
 6. $(A^T A)^{-1}$ exists iff A has full col. rank (if $A^T A$ is invertible, can't solve), $n \gg m \implies$ this will happen.
- $\text{Dim}(A) = n \times m$, $\text{Dim}(x) = m \times 1$, $\text{Dim}(b) = n \times 1$
 - Generally $n > m$
 - **Claim:** $A^T A$ is invertible iff A has linearly independent columns.
 - **Proof:** Show $\text{Null}(A^T A) = \text{Null}(A)$
 - $A^T A$ guaranteed to be square $m \times m$
 - Forward: $\vec{q} \in \text{Null}(A), \exists \vec{q} \neq 0 \implies \vec{q} \in \text{Null}(A^T A)$
 - * Because $A\vec{q} = 0 \implies A^T A\vec{q} = A^T 0 = 0$
 - Backward (Converse): $\vec{r} \in \text{Null}(A^T A), \exists \vec{r} \neq 0 \implies \vec{r} \in \text{Null}(A)$
 - * Because $A^T A\vec{r} = 0 \implies \vec{r}^T A^T A\vec{r} = \langle A\vec{r}, A\vec{r} \rangle = \|A\vec{r}\|^2 = \vec{r}^T 0 = 0$ (**Transpose of a dot product reverses the order of the operands in the product**)
 - * $\|A\vec{r}\|^2 = 0 \implies A\vec{r} = 0 \implies \vec{r} \in \text{Null}(A)$
 - Proves $\text{Null}(A^T A)$ and $\text{Null}(A)$ have the same null space.
 - Only vector in null space of $\text{Null}(A^T A)$ is $\vec{0} \implies$ columns of A are linearly independent.

Inverse Formula for 2×2 Matrix:

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \quad (1)$$

Line of Best Fit

- Example: $t_1 = 0, t_2 = 1, t_3 = 2$ and $y(t) = -1, y(t) = -3, y(t) = 1$, Find $y(t) = x_1 t + x_2$
- Set up $A\vec{x} = y(\vec{t}_n)$
- Solve for $A^T A$

Line of Best Fit Example

$$\begin{bmatrix} 0 & 1 \\ 1 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} y(t_1) = -1 \\ y(t_2) = -3 \\ y(t_3) = 1 \end{bmatrix} \quad (2)$$

Find $A^T A$

$$\begin{bmatrix} 0 & 1 & 2 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 1 \\ 2 & 1 \end{bmatrix} = \begin{bmatrix} 5 & 3 \\ 3 & 3 \end{bmatrix} \quad (3)$$

Find $(A^T A)^{-1}$

$$\frac{1}{6} \begin{bmatrix} 3 & -3 \\ -3 & 5 \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ -\frac{1}{2} & \frac{5}{6} \end{bmatrix} \quad (4)$$

Find $x = (A^T A)^{-1} A^T b$

$$\begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ -\frac{1}{2} & \frac{5}{6} \end{bmatrix} \begin{bmatrix} 0 & 1 & 2 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} -1 \\ -3 \\ 1 \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} & 0 & \frac{1}{2} \\ \frac{2}{3} & \frac{1}{3} & \frac{1}{6} \end{bmatrix} \begin{bmatrix} -1 \\ -3 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \end{bmatrix} \quad (5)$$

Plug in to solve for $y(t)$

$$y(t) = x_1 t + x_2 = t - 2 \quad (6)$$

Find $\varepsilon = \vec{b} - A\vec{x}$

$$\varepsilon = \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} \quad (7)$$

Determine $\|\varepsilon\|^2$

$$\|\varepsilon\|^2 = 1^2 + (-2)^2 + 1^2 = 6 \quad (8)$$

Least Squares Calculus Argument

1. Smallest $\|\varepsilon\|^2$ is the sum of the squares of the distance between each actual measurement (point) and the line of best fit.

Least Squares Calculus:

$$\|\varepsilon\|^2 = \sum_{i=1}^{n(=3)} \varepsilon^2 = \sum_{i=1}^{n(=3)} (x_1 t_i + x_2 - y(t_i))^2 = (x_2 + 1)^2 + (x_1 + x_2 + 3)^2 + (2x_1 + x_2 - 1)^2 \quad (9)$$

Find minimum w.r.t x_1 :

$$\frac{\partial \|\varepsilon\|}{\partial x_1} = 0 + 2(x_1 + x_2 + 3) + 4(2x_1 + x_2 - 1) = 0 \implies 10x_1 + 6x_2 = -2 \quad (10)$$

Find minimum w.r.t x_2 :

$$\frac{\partial \|\varepsilon\|}{\partial x_2} = 2(x_2 + 1) + 2(x_1 + x_2 + 3) + 2(2x_1 + x_2 - 1) = 0 \implies x_1 + x_2 = -1 \quad (11)$$

Solve for x_1, x_2 :

$$x_1 = -x_2 - 1, 10(-x_2 - 1) + 6x_2 = -2, x_2 = -2, x_1 = 1 \quad (12)$$