

The Method of the Least Square

Prof. Dr. Nobuyuki TOSE

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 $\text{Im}(A)$ is the subspace spanned (generated) by $\vec{a}_1, \cdots, \vec{a}_n$

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- **Theorem** $(A\vec{x}, \vec{y}) = (\vec{x}, {}^tA\vec{y})$

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- **Problem** Find the minimum value of $\|\vec{e} - A\vec{v}\| = \|\vec{e} - s\vec{a}_1 - t\vec{a}_2\|$

The Key Idea

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when $A\vec{v}$ is the orthogonal projection of $\vec{e}$ on V .

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- $\vec{w} \in \text{Im}(A) \rightarrow \vec{w} = A\vec{v}$
- $(\vec{e} - A\vec{v}) \perp \text{Im}(A)$
 $\Leftrightarrow (\vec{e} - A\vec{v}, A\vec{\beta}) = 0 \quad (\vec{\beta} \in \mathbb{R}^2)$
 $\Leftrightarrow ({}^t A (\vec{e} - A\vec{v}), \vec{\beta}) = 0 \quad (\vec{\beta} \in \mathbb{R}^2)$
 $\Leftrightarrow {}^t A (\vec{e} - A\vec{v}) = \vec{0}$
 $\Leftrightarrow {}^t A A \vec{v} = {}^t A \vec{e}$

Method of the Least Square No2

- Orth. Proj. $\vec{w} = A\vec{v}$

Method of the Least Square No2

- Orth. Proj. $\vec{w} = A\vec{v}$
- ${}^tAA\vec{v} = {}^tA\vec{e}$

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$${}^tAA = \begin{pmatrix} 0 & 1 & 1 \\ 2 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 2 \\ 1 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 1 & 5 \end{pmatrix}$$

Method of the Least Square No2

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- ${}^tA\vec{e} = \begin{pmatrix} 0 & 1 & 1 \\ 2 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 2 \end{pmatrix}$

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- $\begin{pmatrix} 2 & 1 \\ 1 & 5 \end{pmatrix} \begin{pmatrix} s \\ t \end{pmatrix} = \begin{pmatrix} 0 \\ 2 \end{pmatrix}$

Method of the Least Square No2

- Orth. Proj. $\vec{w} = A\vec{v}$
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- $\begin{pmatrix} 2 & 1 \\ 1 & 5 \end{pmatrix} \begin{pmatrix} s \\ t \end{pmatrix} = \begin{pmatrix} 0 \\ 2 \end{pmatrix} \rightarrow \begin{pmatrix} s \\ t \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 1 & 5 \end{pmatrix}^{-1} \begin{pmatrix} 0 \\ 2 \end{pmatrix}$

Method of the Least Square No2

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- $\begin{pmatrix} 2 & 1 \\ 1 & 5 \end{pmatrix} \begin{pmatrix} s \\ t \end{pmatrix} = \begin{pmatrix} 0 \\ 2 \end{pmatrix} \rightarrow \begin{pmatrix} s \\ t \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 1 & 5 \end{pmatrix}^{-1} \begin{pmatrix} 0 \\ 2 \end{pmatrix}$
 $= \frac{1}{9} \begin{pmatrix} 5 & -1 \\ -1 & 2 \end{pmatrix} \begin{pmatrix} 0 \\ 2 \end{pmatrix} = \frac{1}{9} \begin{pmatrix} -2 \\ 4 \end{pmatrix}$

Method of the Least Square No2

- Orth. Proj. $\vec{w} = A\vec{v}$
- ${}^tAA\vec{v} = {}^tA\vec{e}$

$${}^tAA = \begin{pmatrix} 0 & 1 & 1 \\ 2 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 2 \\ 1 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 1 & 5 \end{pmatrix}$$

- ${}^tA\vec{e} = \begin{pmatrix} 0 & 1 & 1 \\ 2 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 2 \end{pmatrix}$

- $\begin{pmatrix} 2 & 1 \\ 1 & 5 \end{pmatrix} \begin{pmatrix} s \\ t \end{pmatrix} = \begin{pmatrix} 0 \\ 2 \end{pmatrix} \rightarrow \begin{pmatrix} s \\ t \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 1 & 5 \end{pmatrix}^{-1} \begin{pmatrix} 0 \\ 2 \end{pmatrix}$
 $= \frac{1}{9} \begin{pmatrix} 5 & -1 \\ -1 & 2 \end{pmatrix} \begin{pmatrix} 0 \\ 2 \end{pmatrix} = \frac{1}{9} \begin{pmatrix} -2 \\ 4 \end{pmatrix}$