

Solution of Linear Systems

Application: To be used for calculation of linear transformation based on sets of landmarks, e.g.

Materials and Matlab

- <http://audition.ens.fr/brette/calculscientifique/lecture6.pdf>
- http://en.wikipedia.org/wiki/Overdetermined_system
- <http://www.mathworks.com/help/toolbox/optim/ug/brhkghv-18.html>
- <http://www.mathworks.com/help/techdoc/math/f4-2224.html#f4-2282>

Linear Systems

$$A x = b$$

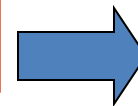
Square system:

- unique solution
- Gaussian elimination

$$A x = b$$

Rectangular system ??

- underconstrained:
infinity of solutions
- overconstrained:
no solution



Minimize $|Ax-b|$

How do you solve overconstrained linear equations??

- Define $E = |\mathbf{e}|^2 = \mathbf{e} \cdot \mathbf{e}$ with

$$\begin{aligned}\mathbf{e} &= A\mathbf{x} - \mathbf{b} = \left[\begin{array}{c|c|c|c} \mathbf{c}_1 & \mathbf{c}_2 & \dots & \mathbf{c}_n \end{array} \right] \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} - \mathbf{b} \\ &= x_1\mathbf{c}_1 + x_2\mathbf{c}_2 + \dots + x_n\mathbf{c}_n - \mathbf{b}\end{aligned}$$

- At a minimum,

$$\begin{aligned}\frac{\partial E}{\partial x_i} &= \frac{\partial \mathbf{e}}{\partial x_i} \cdot \mathbf{e} + \mathbf{e} \cdot \frac{\partial \mathbf{e}}{\partial x_i} = 2 \frac{\partial \mathbf{e}}{\partial x_i} \cdot \mathbf{e} \\ &= 2 \frac{\partial}{\partial x_i} (x_1\mathbf{c}_1 + \dots + x_n\mathbf{c}_n - \mathbf{b}) \cdot \mathbf{e} = 2\mathbf{c}_i \cdot \mathbf{e} \\ &= 2\mathbf{c}_i^T (A\mathbf{x} - \mathbf{b}) = 0\end{aligned}$$

- or

$$0 = \begin{bmatrix} \mathbf{c}_i^T \\ \vdots \\ \mathbf{c}_n^T \end{bmatrix} (A\mathbf{x} - \mathbf{b}) = A^T (A\mathbf{x} - \mathbf{b}) \Rightarrow A^T A\mathbf{x} = A^T \mathbf{b},$$

where $\mathbf{x} = A^\dagger \mathbf{b}$ and $A^\dagger = (A^T A)^{-1} A^T$ is the *pseudoinverse* of A !

Overconstrained Problems in Matlab

Problem: solve A for $X^*A=Y$:

- if full rank: unique solution
- if overconstrained: This means we want to find the solution that minimizes $\sum_{(x,y) \text{ pairs}} (y-xA)^2$

- **Solution 1:** Left Matrix Divide: $A=X \setminus Y$ is the matrix division of X into Y, which is roughly the same as $INV(X)^*Y$, except it is computed in a different way.
- **Solution 2:** Use pseudoinverse: $X=A^+*Y$. The pseudoinverse is calculated as: $A = Y^*pinv(X)$. However, it may be easier to write out the system as $X^*A=Y$ and then do $A = X \setminus Y$ (solution 1) which is pretty standard.