

Fall 2022 ME424 Modern Control and Estimation

Lecture Note 3
Least Squares and Basic System Identification

Prof. Wei Zhang
Department of Mechanical and Energy Engineering
SUSTech Institute of Robotics
Southern University of Science and Technology

`zhangw3@sustech.edu.cn`
<https://www.wzhanglab.site/>

Outline

- **Least-Squares Problem Formulation**
- Solution to Linear Least-Squares Problems
- Linear Least-Squares Examples
- Applications to System ID
- Nonlinear Least Squares

- Last lecture: obtain discrete-time linear state space model from
 - physical process
 - given continuous time state space model
 - given nonlinear state space model
 - given discrete time transfer function
- **The goal of this lecture note:**
 - learn how to build model based on observed input-output data pairs
 - General case beyond the scope of this course
 - Focus on special case, where first obtain transfer function model from input-output data pairs, and then obtain the corresponding state space model
- **Main method: Least Squares**

Least-Squares Problem Formulation:

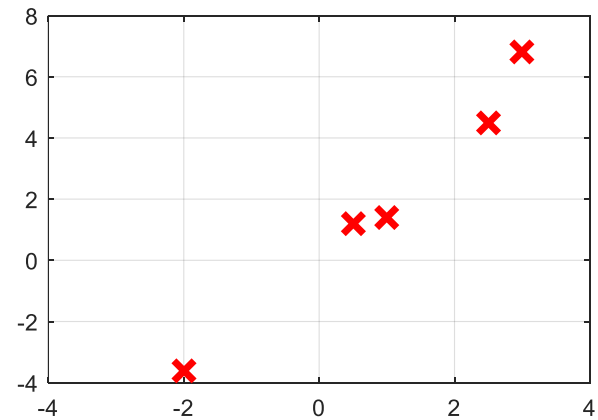
- Measurement Equation:

$$y = g(\theta) + v$$

- $y \in R^m$: measurements data
- $\theta \in \Theta \subseteq R^n$: parameter to be estimated, where Θ is the constraint set for feasible parameters
- $v \in R^m$: unknown measurement noise
- $g: R^n \rightarrow R^m$: known (possibly) nonlinear function relates θ with measurement y

■ Least Squares Example:

i	1	2	3	4	5
x	1	0.5	-2	3	2.5
y	1.4	1.2	-3.6	6.8	4.5



Least-Squares Problem Formulation:

- **Problem Statement:** Find the best parameter in the constraint set Θ that minimizes the difference between the model and the measured data

$$\min_{\theta \in \Theta} J(\theta) = \min_{\theta \in \Theta} \|y - g(\theta)\|^2$$

- **Linear Least Squares:** $g(\theta) = H\theta$, where $H \in R^{m \times n}$ is a given deterministic matrix

Outline

- Least-Squares Problem Formulation
- **Solution to Linear Least-Squares Problems**
- Linear Least-Squares Examples
- Applications to System ID
- Nonlinear Least Squares

Optimization of multivariable function

- 1st –order necessary condition for optimality of $J(\theta)$

- Matrix calculus:

- If $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$, then $\frac{\partial f}{\partial x}(x) = Df(x) =$

Optimization of multivariable function

- **Gradient:** For scalar valued multivariate function $f: R^n \rightarrow R$, its gradient is defined as:

- For $f: R^n \rightarrow R$, notational convention $\nabla f(x) = \left(\frac{\partial f}{\partial x}(x) \right)^T$

- Some references use $\frac{\partial f}{\partial x}$ to denote gradient

- Directional derivative: $Df(x; d) = \lim_{\alpha \rightarrow 0} \frac{f(x+\alpha d) - f(x)}{h}$

- Some calculus examples:

- $f(x) = Ax$

- $f(x) = x^T A x$

- **Exercise:** compute $\frac{\partial f}{\partial x}(x)$, where $x \in R^n$, and $f(x) = x^T x \cdot x$

- Derivation of linear least square solutions
 - Normal equation:

- Solution with full rank H

- Geometric interpretation of linear least squares

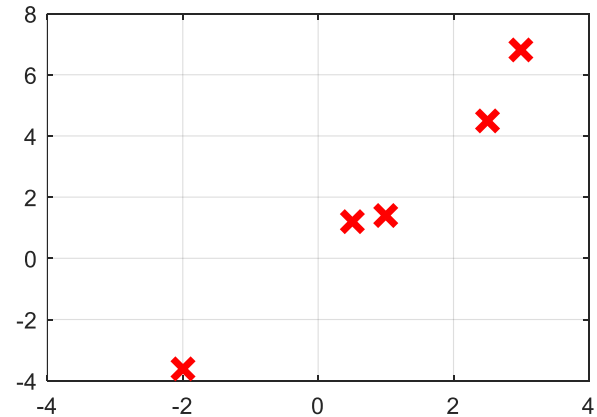
Outline

- Least-Squares Problem Formulation
- Solution to Linear Least-Squares Problems
- **Linear Least-Squares Examples**
- Applications to System ID
- Nonlinear Least Squares

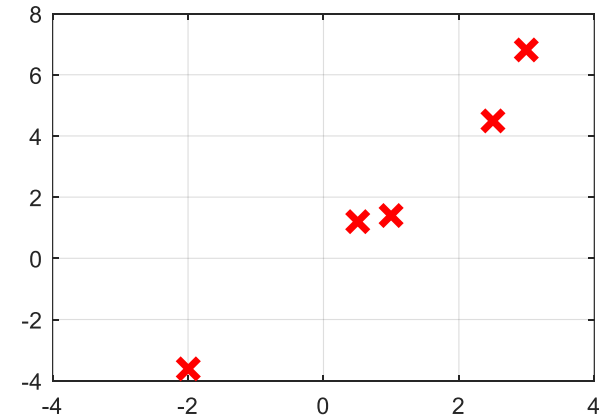
Linear Least Squares Example:

i	1	2	3	4	5
x	1	0.5	-2	3	2.5
y	1.4	1.2	-3.6	6.8	4.5

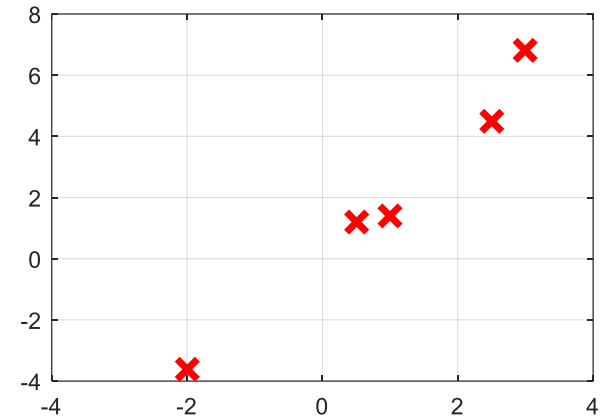
- Assume $y = \alpha x + \beta$



- Change hypothesis, assume $y = be^{ax}$



- Change hypothesis, assume that $y = \alpha_0 + \alpha_1 x + \alpha_2 x^2$



Outline

- Least-Squares Problem Formulation
- Solution to Linear Least-Squares Problems
- Linear Least-Squares Examples
- **Applications to System ID**
- Nonlinear Least Squares

Application to System Identification for Linear Systems

- ARX(p, q) model :(Autoregressive with exogenous input)

$$y(k) + \alpha_1 y(k-1) + \dots + \alpha_p y(k-p) \\ = \beta_0 u(k) + \beta_1 u(k-1) + \dots + \beta_q u(k-q) + v(k)$$

- $v(k)$: noise signal

- Model parameter: $\theta = [\alpha_1, \dots, \alpha_p, \beta_0, \beta_1, \dots, \beta_q]^T$

- One-step predictor:

$$\hat{y}(k|\theta) = -\alpha_1 y(k-1) - \dots - \alpha_p y(k-p) \\ + \beta_0 u(k) + \beta_1 u(k-1) + \dots + \beta_q u(k-q)$$

- System ID problem for ARX model:

Given data pairs $\{(u(k), y(k))\}_{k \leq N}$, find the parameter vector θ that minimizes cost:

- $J(\hat{\theta}) = \sum_{k=1}^N \|\hat{y}(k|\hat{\theta}) - y(k)\|^2$

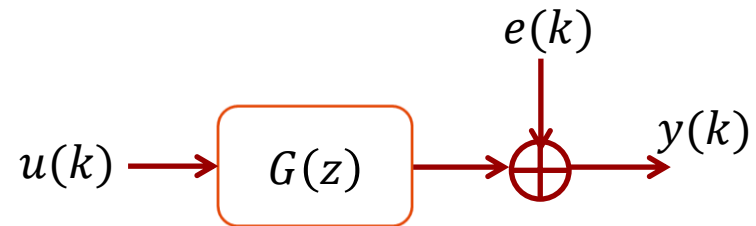
- Formulate as least square problem:
given data set, $(u_1, y_1), (u_2, y_2), \dots, (u_m, y_m)$
- Regressor:

- Derivation continued

■ System ID Example I:

$G(z) = \frac{(z^2+b)}{z^3+az}$, find best estimate for a, b ,
given data set $(u_1, y_1), (u_2, y_2), \dots, (u_{20}, y_{20})$

■ System ID Example 2:



- $G(z) = \frac{z-1}{z-a}$, where a is an unknown scalar
- Data: $u(1) = 1, u(2) = \frac{1}{2}, u(3) = 1, y(1) = 2, y(2) = 1, y(3) = 2$

Outline

- Least-Squares Problem Formulation
- Solution to Linear Least-Squares Problems
- Linear Least-Squares Examples
- Applications to System ID
- **Nonlinear Least Squares**

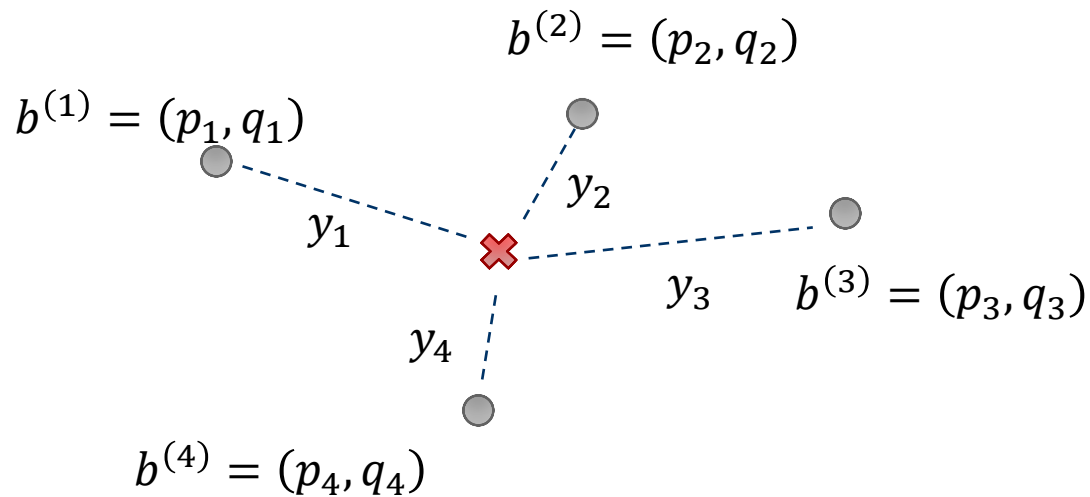
- Nonlinear Least Squares:



$$\min_{\theta \in \Theta} J(\theta) = \|y - g(\theta)\|^2$$

- For general nonlinear function $g(\theta)$, analytical solution to the above optimization is not available
- Numerical optimization algorithms can be used to find the optimizer

$$\theta^* = \operatorname{argmin}_{\theta \in \Theta} J(\theta)$$

- Nonlinear Least Square Example: Navigation by range measurement:



-  : beacons with known positions $b^{(i)} = (p_i, q_i)$
-  : target with unknown position $\theta = (\theta_1, \theta_2)$
- y_i : known measured distance or range from beacon i :
typical assumption: $y_i = \|b^{(i)} - \theta\| + v_i$

- Given measurements y_1, y_2, \dots, y_m , find the best target location θ
- We can choose cost function: $J(\theta) = \sum_{i=1}^m \left(y_i - \|b^{(i)} - \theta\| \right)^2$
- Coding Example