

## Behavioral Approach for Exact and Approximate Modeling

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I. Markovsky, etc. "Exact and Approximate Modeling of Linear Systems -A Behavioral Approach"



## Outline

- Introduction
- Behavioral Approach
- Definitions
- Examples

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## Introduction

- Modeling: use something simpler, well understood, replicable, applicable to represent/mimic/explain something complex, mysterious, autonomous.
- Key points:
  - Data: by disturbance and observation
  - Constraints: knowledge about the system
  - Evaluation: misfit, latency



## **Behavioral Approach**

- Input-State-Output Approach
  - Classify data into input and output
  - Represent the model by defining the relation between input and output
    - For example: transfer function
- Behavioral Approach
  - Group all data as observation
  - Represent the model using observed data
    - For example: data matrix



## **Definitions**

- U: Universion of all possible outcomes
- D: Observed data
- B: model

# A model B is a subset of U which explains D $D \subseteq B \subseteq U$

 $B_1$  is more powerful than  $B_2$  if

$$D \subseteq B_1 \subset B_2 \subseteq U$$

Due to noise or approximation, B may not explain D Assume D<sup>B</sup> is the data predicted by B

• M: misfit  $M(D,B) = \|M(d_1,d_1^B),...\|$ 

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## Example

- We have data D consisting of 2 variables and 10 data points
- We want a linear model that passes through origin (0,0) which minimize the misfit M





## **Misfit**

- We can define misfit function M(D,B)
- If we define it as 2-norm

$$\min_{B} \{ \min_{D^{B}} \{ \| D - D^{B} \|_{2}, D^{B} \in B \}, (0,0) \in B \}$$

It is total-least-square (TLS) problem





## **Solution**

Let  $D = U \sum V^T$  be ranking singular value decomposition

$$U = \begin{bmatrix} U_1 & U_2 \end{bmatrix} \qquad \sum = \begin{bmatrix} \sum_1 & 0 \\ 0 & \sum_2 \end{bmatrix} \qquad V = \begin{bmatrix} V_1 & V_2 \end{bmatrix}$$

Then we can have  $D^B = U_1 \sum_1 V_1^T$ 



## **Misfit**

- We can define misfit function M(D,B)
- If we define it as weighted 2-norm

 $\min_{B} \{ \min_{D^{B}} \{ \left\| W_{L}(D - D^{B}) W_{R} \right\|_{2}, D^{B} \in B \}, (0, 0) \in B \}$ 

It becomes a generalized
TLS

Let  $D_m$  be the modified data matrix  $D_m = \sqrt{w_l} D \sqrt{w_r}$ Then the problem is transformed into TLS on  $D_m$ 



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