

# 21. Multibody Parameter Estimation

Physics based reasoning and do physics inspired neural networks,  
Physics inspired neutral networks.

## Today's Topic outline

- Goal: generate plant model for control
- In real world, if we know the multibody equation, we can do way better. Zero shot

### 1. Dynamics of Throwing

Estimating the parameters determines take off conditions

### 2. Parameters to identify

Separate kinematics & dynamics

$$\alpha = [m, h, \sum]' \quad 1 + 3 + 6 \text{ for each link}$$

### 3. Lesson 1 adds additional constraints

Positive semi definite matrix: pseudo inertia

Two approaches:

- Semi Definite Programming
- Gradient Decent Programming

Break pseudo matrix into products of uptriangular matrix,  $J = U \cdot U'$

### 4. Lesson 2 Equation Error VS. Simulation Error

- Equation error is one step error, reset at every step, no compounding accumulation
- Simulation error is long term prediction with compounding error

### 5. Lesson 3 Use Inverse Dynamics (Not Forward Dynamics)

Multibody dynamics equation is convex in inertia parameters. Write dynamics in linear in lumped parameters.

## 6. Lesson 4 not all multibody quantities are identifiable

- Idea 1: optimize only identifiable parameters
- Idea 2: add regularizer
- Idea 3: optimal experiment design
- Idea 4: online parameter identification, i.e. Adaptive Control, the Controller takes in the lumped parameter.

book:<Applied Nonlinear Control>, Jean-Jacques