

# Smooth discrete feedback control of walking robots

An intermediate between fully passive and high bandwidth feedback control

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(work with Andy Ruina)

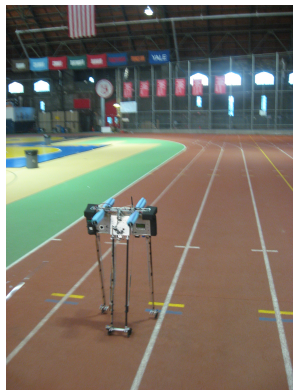
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# What is interesting about this control architecture?



Cornell Ranger walked 14.3 miles (=23 kms.)  
using only 3 cents worth of electricity!

# State of the art in walking robots

## High Bandwidth Feedback Control

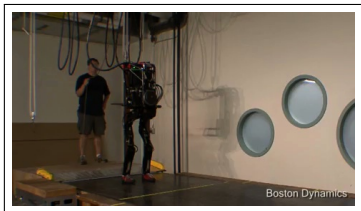


Figure: PETMAN from Boston Dynamics

- Robust, Versatile.
- Energy Inefficient.

## Passive Dynamic Walkers

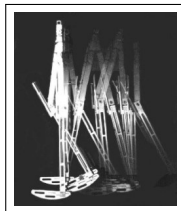


Figure: Passive Knead Walker

- Energy efficient
- Not Robust.

# Is there an intermediate approach?

- can be made close to optimal feedback control.
- can be made robust.
- can be made dumb.

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3. Find open loop approximate optimal control policy (design choice)
  - ▶ Torques as a function of state.
  - ▶ Maximize simplicity of representation.
  - ▶ Minimize number of parameters.



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  - ▶ Maximize simplicity of representation.
  - ▶ Minimize number of parameters.
4. Stabilize the approximate optimal control policy (details ahead)

# Stability is easy

Equation of motion

$$\delta x_{n+1} = A\delta x_n + B\delta u_n$$

Output equation

$$\delta z_{n+1} = C\delta x_n + D\delta u_n$$

- $n = \text{step number}^*$
- $\delta x_n =$  differential about nominal value for state vector.
- $\delta u_n =$  differential about nominal value for control vector.
- $\delta z_n =$  differential about nominal value for output vector.
- $A = \text{Jacobian} = \partial(\delta x_{n+1})/\partial(\delta x_n)$
- $B = \text{Sensitivity} = \partial(\delta x_{n+1})/\partial(\delta u_n)$
- $C = \text{Jacobian} = \partial(\delta z_{n+1})/\partial(\delta x_n)$
- $D = \text{Sensitivity} = \partial(\delta z_{n+1})/\partial(\delta u_n)$

\* could be time or state based

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## Output Equation

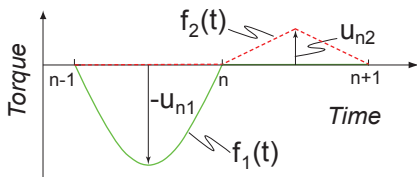
$$\delta z_{n+1} = C\delta x_n + D\delta u_n$$

e.g.

$$\delta u_n = \begin{bmatrix} u_{n1} f_1(t) \\ u_{n2} f_2(t) \end{bmatrix}$$

## NOTE:

- $\delta u_n$  is intermittent.
- $\delta u_n$  can be made smooth.



# Stability is easy

How to use the output equation for control?

For linear control,

$$\delta u_n = K \delta x_n$$

Thus,

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e.g. dead beat control

$$(\delta z_{n+1} = 0)$$

$$\delta u_n = K \delta x_n = -D^{-1} C \delta x_n$$



# Example 1: Inverted Pendulum

Linearization over 1 sec.

i.e.  $n$  (seconds) = 1,2,3 ...

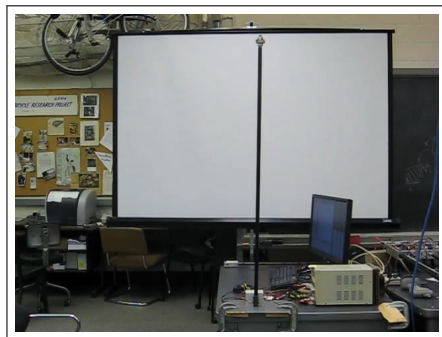
## Output Equation

$$\overbrace{\delta x_{n+1}}^{\text{state}} = C\delta x_n + D \underbrace{K\delta x_n}_{\delta u_n}$$

For dead beat control

$$\delta x_{n+1} = 0$$

$$\delta u_n = K\delta x_n = -D^{-1}C\delta x_n$$



## Example 2: Walking Robot

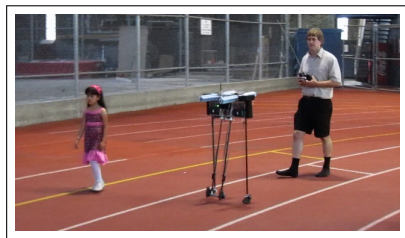
Linearization about upright position.  
and  $n$  (step number) = 1,2,3 ...

### Output Equation

$$\underbrace{\delta E_{n+1}}_{\text{Mech. Energy}} = F\delta x_n + G \underbrace{K\delta x_n}_{\delta u_n} \quad \text{Hip Torque}$$

For dead beat control  $\delta E_{n+1} = 0$

$$\delta u_n = K\delta x_n = -D^{-1}C\delta x_n$$



## Walk Statistics

Distance:

14.3 mi (=23 km).

Time:

10 hrs, 40 min, 48 sec

Speed:

1.34 mi/hr (=2.15 km/hr)

No. of Steps:

65,185.

Power:

24.5 watt

Energy:

262 watt hours

$\text{COT} = \text{Energy} / (\text{Distance} \times \text{Weight})$ :

0.49

# Thank You

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