

# CS 4649/7649

## Robot Intelligence: Planning

### Linearized Control

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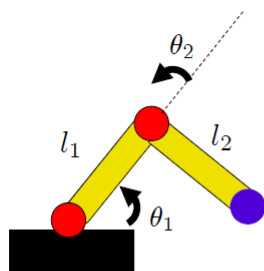
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11/06/2014

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\*Slides based in part on Dr. Mike Stilman's slides

## Manipulation: Linear Dynamics?



2-DOF Robot Arm

$$\mathbf{T}_1^0 = \begin{bmatrix} c_1 & -s_1 & 0 \\ s_1 & c_1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{T}_2^1 = \begin{bmatrix} c_2 & -s_2 & l_1 \\ s_2 & c_2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{T}_3^2 = \begin{bmatrix} 1 & 0 & l_2 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{T}_2^0 = \mathbf{T}_1^0 \mathbf{T}_2^1 = \begin{bmatrix} c_1 c_2 - s_1 s_2 & -c_1 s_2 - s_1 c_2 & l_1 c_1 \\ s_1 c_2 + c_1 s_2 & -s_1 s_2 + c_1 c_2 & l_1 s_1 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} c_{12} & -s_{12} & l_1 c_1 \\ s_{12} & c_{12} & l_1 s_1 \\ 0 & 0 & 1 \end{bmatrix}$$

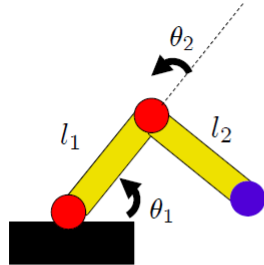
$$\mathbf{T}_3^0 = \mathbf{T}_2^0 \mathbf{T}_3^2 = \begin{bmatrix} c_{12} & -s_{12} & l_1 c_1 + l_2 c_{12} \\ s_{12} & c_{12} & l_1 s_1 + l_2 s_{12} \\ 0 & 0 & 1 \end{bmatrix}$$

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## Manipulation: Linear Dynamics?



2-DOF Robot Arm

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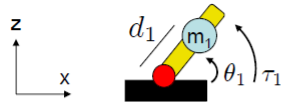
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$$\mathbf{T}_2^0 = \mathbf{T}_1^0 \mathbf{T}_2^1 = \begin{bmatrix} c_1 c_2 - s_1 s_2 & -c_1 s_2 - s_1 c_2 & l_1 c_1 \\ s_1 c_2 + c_1 s_2 & -s_1 s_2 + c_1 c_2 & l_1 s_1 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} c_{12} & -s_{12} & l_1 c_1 \\ s_{12} & c_{12} & l_1 s_1 \\ 0 & 0 & 1 \end{bmatrix}$$

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## Manipulation: Gravity Compensation



$$x_1 = d_1 \cos(\theta_1)$$

Torque due to gravity:

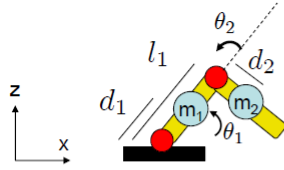
$$\tau_g = -d_1 \cos(\theta_1) m_1 g$$

$$\tau_1 = -d_1 \cos(\theta_1) m_1 g + \tau_1^a$$

$$\tau_1^a = -K_p(\theta_1 - \theta_{1d}) - K_d \dot{\theta}_1$$

$$\text{Better: } \tau_1^a = d_1 c_1 m_1 g - K_p(\theta_1 - \theta_{1d}) - K_d \dot{\theta}_1$$

## Manipulation: Gravity Compensation



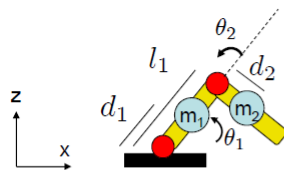
$$\begin{aligned}x_1 &= d_1 c_1 \\x_2 &= l_1 c_1 + d_2 c_2\end{aligned}$$

Torque Due to Gravity:

$$\begin{aligned}\tau_2^g &= -d_2 c_{12} m_2 g \\ \tau_1^g &= -d_1 c_1 m_1 g - (l_1 c_1 + d_2 c_{12}) m_2 g\end{aligned}$$

How do we do Position Control?

## Manipulation: Gravity Compensation



$$\begin{aligned}x_1 &= d_1 c_1 \\x_2 &= l_1 c_1 + d_2 c_2\end{aligned}$$

Torque Due to Gravity:

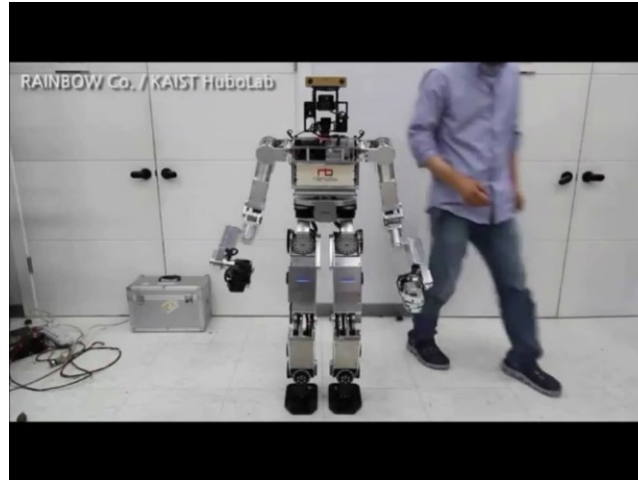
$$\begin{aligned}\tau_2^g &= -d_2 c_{12} m_2 g \\ \tau_1^g &= -d_1 c_1 m_1 g - (l_1 c_1 + d_2 c_{12}) m_2 g\end{aligned}$$

Position Control with Gravity Compensation:

$$\begin{bmatrix} \tau_1 \\ \tau_2 \end{bmatrix} = \begin{bmatrix} d_2 c_{12} m_2 \\ d_1 c_1 m_1 + (l_1 c_1 + d_2 c_{12}) m_2 \end{bmatrix} g - \mathbf{K}_p \begin{bmatrix} \theta_1 - \theta_{1d} \\ \theta_2 - \theta_{2d} \end{bmatrix} - \mathbf{K}_d \begin{bmatrix} \dot{\theta}_1 \\ \dot{\theta}_2 \end{bmatrix}$$

Gains are typically diagonal matrices

## Manipulation: Gravity & Position



## Manipulation: Full Model Control

$$M(\mathbf{q})\ddot{\mathbf{q}} + C(\mathbf{q}, \dot{\mathbf{q}})\dot{\mathbf{q}} + \mathbf{G}(\mathbf{q}) = \boldsymbol{\tau}$$

Augmented PD:

$$M(\mathbf{q})\ddot{\mathbf{q}}_d + C(\mathbf{q}, \dot{\mathbf{q}})\dot{\mathbf{q}}_d + \mathbf{G}(\mathbf{q}) - \mathbf{K}_d(\dot{\mathbf{q}} - \dot{\mathbf{q}}_d) - \mathbf{K}_p(\mathbf{q} - \mathbf{q}_d) = \boldsymbol{\tau}$$

Computed Torque:

$$M(\mathbf{q})(\ddot{\mathbf{q}}_d - \mathbf{K}_d(\dot{\mathbf{q}} - \dot{\mathbf{q}}_d) - \mathbf{K}_p(\mathbf{q} - \mathbf{q}_d)) + C(\mathbf{q}, \dot{\mathbf{q}})\dot{\mathbf{q}} + \mathbf{G}(\mathbf{q}) = \boldsymbol{\tau}$$

## Joint Space to Work Space

$$\delta \mathbf{x} = \mathbf{J} \delta \mathbf{q} \quad \mathbf{W} = \mathbf{F}^T \mathbf{x} \quad \delta \mathbf{W} = \mathbf{F}^T \delta \mathbf{x} \quad \delta \mathbf{W} = \boldsymbol{\tau}^T \delta \mathbf{q}$$

$$\mathbf{F}^T \delta \mathbf{x} = \boldsymbol{\tau}^T \delta \mathbf{q}$$

$$\mathbf{F}^T \mathbf{J} \delta \mathbf{q} = \boldsymbol{\tau}^T \delta \mathbf{q}$$

Open Loop Force Control

$$\boldsymbol{\tau} = \mathbf{J}^T \mathbf{F}$$

Velocity Jacobian for Link Mass i:  $\mathbf{J}_{vi} \quad \mathbf{J}_{vi}^T \mathbf{g} = ?$

$$\mathbf{g} = \text{gravity vector} = [0 \quad 0 \quad 9.8]^T$$

## Joint Space to Work Space

$$\delta \mathbf{x} = \mathbf{J} \delta \mathbf{q} \quad \mathbf{W} = \mathbf{F}^T \mathbf{x} \quad \delta \mathbf{W} = \mathbf{F}^T \delta \mathbf{x} \quad \delta \mathbf{W} = \boldsymbol{\tau}^T \delta \mathbf{q}$$

$$\mathbf{F}^T \delta \mathbf{x} = \boldsymbol{\tau}^T \delta \mathbf{q}$$

$$\mathbf{F}^T \mathbf{J} \delta \mathbf{q} = \boldsymbol{\tau}^T \delta \mathbf{q}$$

Open Loop Force Control

$$\boldsymbol{\tau} = \mathbf{J}^T \mathbf{F}$$

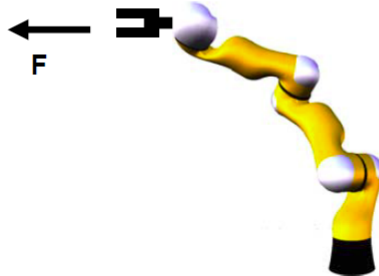
Velocity Jacobian for Link Mass i:  $\mathbf{J}_{vi} \quad \mathbf{J}_{vi}^T \mathbf{g} = ?$

$$\mathbf{G} = -(\mathbf{J}_{v1}^T (m_1 \mathbf{g}) + \mathbf{J}_{v2}^T (m_2 \mathbf{g}) \dots \mathbf{J}_{vn}^T (m_n \mathbf{g}))$$

$$\boldsymbol{\tau} = \mathbf{G}(\mathbf{q}) + \mathbf{J}^T \mathbf{F}$$

## Manipulation: Workspace Control

$$\begin{aligned}
 M(\mathbf{q})\ddot{\mathbf{q}} + C(\mathbf{q}, \dot{\mathbf{q}})\dot{\mathbf{q}} + \mathbf{G}(\mathbf{q}) &= \boldsymbol{\tau} & \bar{M} &= J^{-T}MJ^T \\
 \bar{M}(\mathbf{q})\ddot{\mathbf{x}} + \bar{C}(\mathbf{q}, \dot{\mathbf{q}})\dot{\mathbf{x}} + \bar{\mathbf{G}}(\mathbf{q}) &= \mathbf{F} & \bar{C} &= J^{-T}(CJ^{-1} + MJ^{-1}) \\
 & & \bar{\mathbf{G}} &= J^{-T}\mathbf{G} \\
 & & \mathbf{F} &= J^{-T}\boldsymbol{\tau}
 \end{aligned}$$



Khatib '80, Murray '94

## Manipulation: Workspace Control

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 M(\mathbf{q})\ddot{\mathbf{q}} + C(\mathbf{q}, \dot{\mathbf{q}})\dot{\mathbf{q}} + \mathbf{G}(\mathbf{q}) &= \boldsymbol{\tau} & \bar{M} &= J^{-T}MJ^T \\
 \bar{M}(\mathbf{q})\ddot{\mathbf{x}} + \bar{C}(\mathbf{q}, \dot{\mathbf{q}})\dot{\mathbf{x}} + \bar{\mathbf{G}}(\mathbf{q}) &= \mathbf{F} & \bar{C} &= J^{-T}(CJ^{-1} + MJ^{-1}) \\
 & & \bar{\mathbf{G}} &= J^{-T}\mathbf{G} \\
 & & \mathbf{F} &= J^{-T}\boldsymbol{\tau}
 \end{aligned}$$

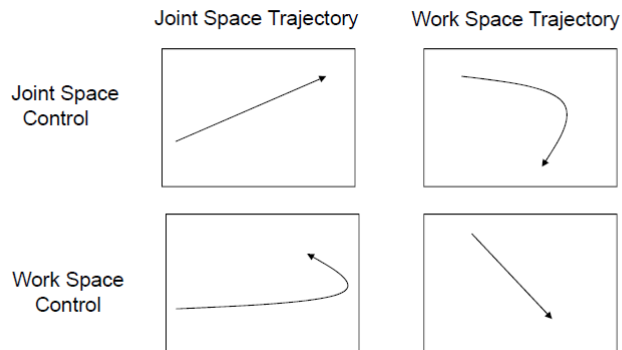
Computed Torque:

$$\begin{aligned}
 \mathbf{F} &= \bar{M}(\mathbf{q})(\ddot{\mathbf{x}}_d - \mathbf{K}_d(\dot{\mathbf{x}} - \dot{\mathbf{x}}_d) - \mathbf{K}_p(\mathbf{x} - \mathbf{x}_d)) + \bar{C}(\mathbf{q}, \dot{\mathbf{q}})\dot{\mathbf{x}} + \bar{\mathbf{G}}(\mathbf{q}) \\
 \boldsymbol{\tau} &= \mathbf{J}^T \mathbf{F}
 \end{aligned}$$

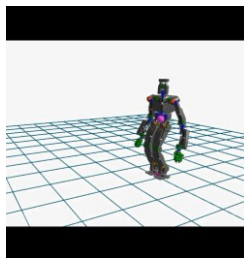
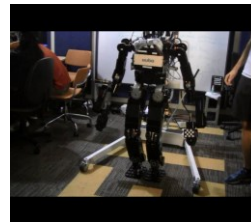
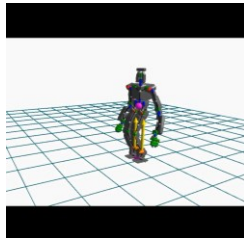
Gravity Augmented PD:

$$\begin{aligned}
 \mathbf{F} &= \bar{\mathbf{G}}(\mathbf{q}) - \mathbf{K}_d(\dot{\mathbf{x}} - \dot{\mathbf{x}}_d) - \mathbf{K}_p(\mathbf{x} - \mathbf{x}_d) \\
 \boldsymbol{\tau} &= \mathbf{J}^T \mathbf{F}
 \end{aligned}$$

## Joint Space vs Work Space



## Locomotion: Walking

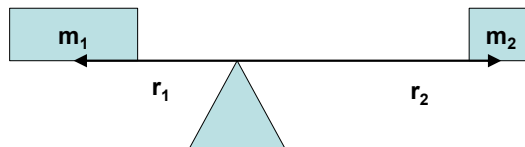


## Balance



## Balance

Moment of force (Torque)



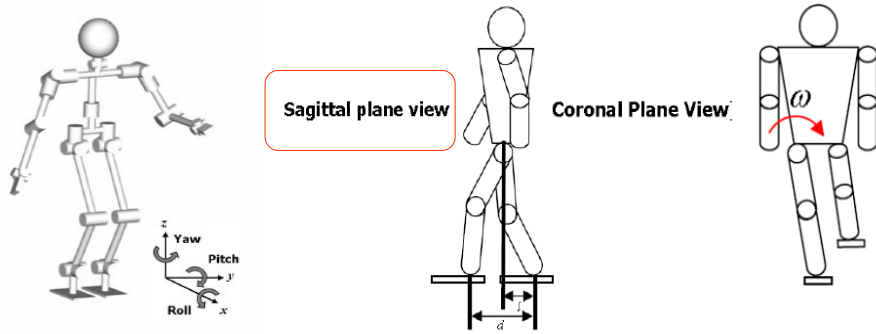
$$\vec{r}_1 \times \vec{F}_1 + \vec{r}_2 \times \vec{F}_2 = 0$$

$$\vec{F}_1 = m_1 \vec{g}$$

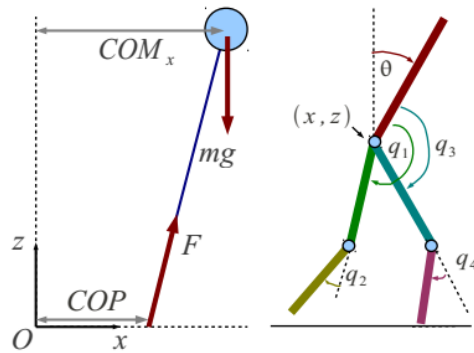
$$\vec{F}_2 = m_2 \vec{g}$$



# Biped Walking

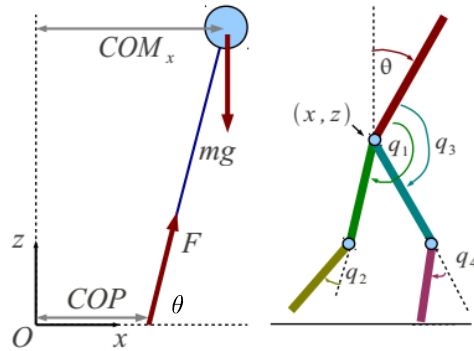


# Inverted Pendulum Model



Center of Mass: COM  
Center of Pressure: COP

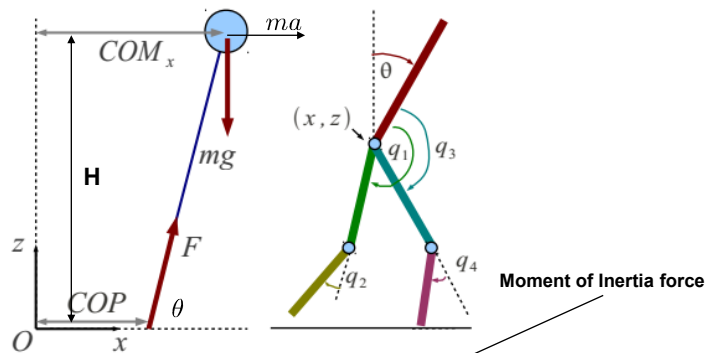
## Inverted Pendulum Model



**If balanced about O**

$$-COM_x mg + COP_x F \sin \theta = 0$$

## Inverted Pendulum Model

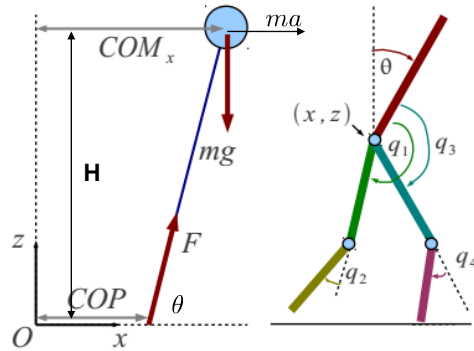


**If the mass is moving with some acceleration in x direction**

$$-x \cdot m \cdot g + p \cdot F \cdot \sin \theta + H \cdot m \cdot a = 0 \quad x = COM_x, p = COP$$

$$p = \frac{x \cdot m \cdot g}{F \cdot \sin \theta} - \frac{H \cdot m \cdot \ddot{x}}{F \cdot \sin \theta} \quad a = \ddot{x}$$

## Inverted Pendulum Model

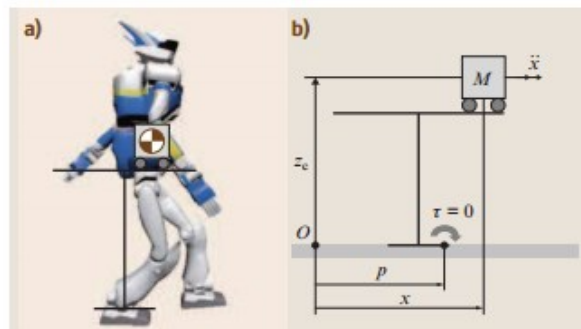


At upright pose,  $F = mg$  &  $\Theta = \pi/2$

$$-x \cdot m \cdot g + p \cdot F \cdot \sin \theta + H \cdot m \cdot a = 0 \quad x = COM_x, \quad p = COP$$

$$p = x - \frac{H}{g} \ddot{x} \quad a = \ddot{x}$$

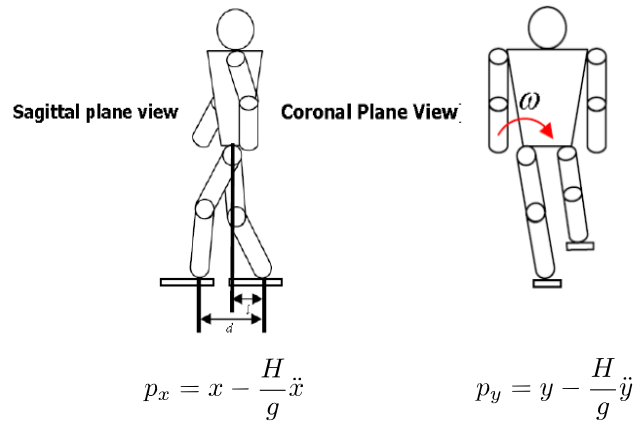
## Table-Cart Model



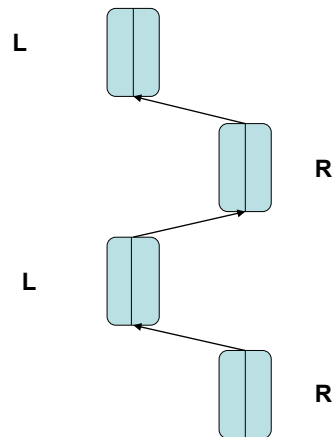
$$p = x - \frac{H}{g} \ddot{x}$$

**P = Zero Moment Point (ZMP)**

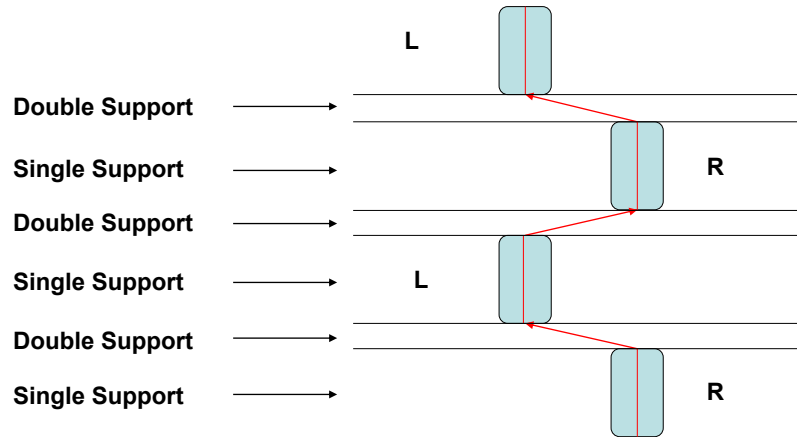
## Table-Cart Model



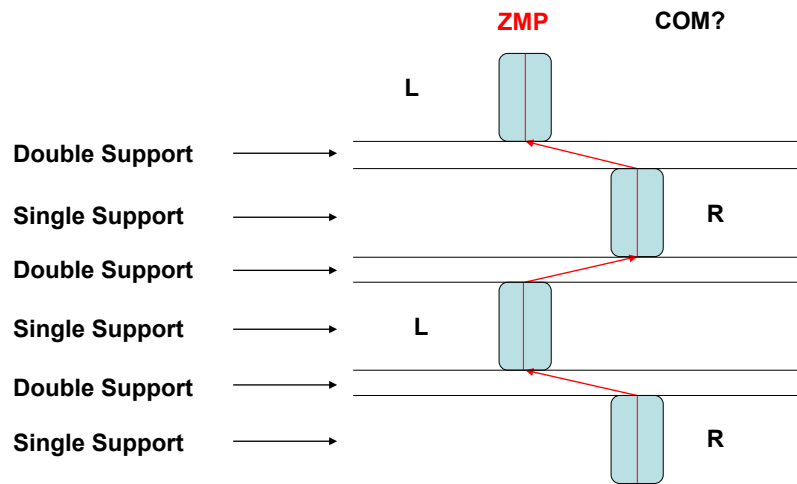
## Footsteps: ZMP vs COM



## Footsteps: ZMP vs COM



## Footsteps: ZMP vs COM



## Linear Optimal Control

$$\mathbf{p}_{zmp} = \mathbf{x}_G - \mathbf{T}_c^2 \ddot{\mathbf{x}}_G \quad T_c = \sqrt{\frac{z_G}{g}}$$

$$\mathbf{p}_{zmp} = [x_{zmp}, y_{zmp}]^T \in \mathbb{R}^2$$

$$\mathbf{x}_G = [x_G, y_G]^T \in \mathbb{R}^2 \quad \frac{d\ddot{\mathbf{x}}_G}{dt} = \mathbf{u} \in \mathbb{R}^2$$

$$\mathbf{X}(t) = [\mathbf{x}^T(t), \dot{\mathbf{x}}^T(t), \ddot{\mathbf{x}}^T(t)]^T \in \mathbb{R}^6$$

$$\mathbf{X}(k+1) = \mathbf{A}\mathbf{X}(k) + \mathbf{B}\mathbf{u}(k)$$

$$\mathbf{y}(k) = \mathbf{p}_{zmp} = \mathbf{C}\mathbf{X}(k) \in \mathbb{R}^2$$

## Linear Optimal Control

$$\mathbf{X}(t) = [\mathbf{x}^T(t), \dot{\mathbf{x}}^T(t), \ddot{\mathbf{x}}^T(t)]^T \in \mathbb{R}^6$$

$$\mathbf{X}(k+1) = \mathbf{A}\mathbf{X}(k) + \mathbf{B}\mathbf{u}(k)$$

$$\mathbf{y}(k) = \mathbf{p}_{zmp} = \mathbf{C}\mathbf{X}(k) \in \mathbb{R}^2$$

$$\mathbf{A} = \begin{bmatrix} \mathbf{I}_2 & \Delta t \mathbf{I}_2 & \frac{\Delta t^2}{2} \mathbf{I}_2 \\ \mathbf{0}_2 & \mathbf{I}_2 & \Delta t \mathbf{I}_2 \\ \mathbf{0}_2 & \mathbf{0}_2 & \mathbf{I}_2 \end{bmatrix} \in \mathbb{R}^{6 \times 6} \quad \mathbf{B} = \begin{bmatrix} \frac{\Delta t^3}{6} \mathbf{I}_2 \\ \frac{\Delta t^2}{2} \mathbf{I}_2 \\ \Delta t \mathbf{I}_2 \end{bmatrix} \in \mathbb{R}^{6 \times 2}$$

$$\mathbf{C} = [\mathbf{I}_2 \quad \mathbf{0}_2 \quad -\mathbf{T}_c^2 \mathbf{I}_2] \in \mathbb{R}^{2 \times 6}$$

## Preview Control

$$\Delta \mathbf{u}(k) = \mathbf{u}(k) - \mathbf{u}(k-1)$$

$$\Delta \mathbf{X}(k) = \mathbf{X}(k) - \mathbf{X}(k-1)$$

$$\Delta \mathbf{y}_d(k) = \mathbf{y}_d(k) - \mathbf{y}_d(k-1)$$

$$\mathbf{y}_d(k+i) = \mathbf{y}_d(k+N) \quad i > N = \text{preview horizon}$$

$$\mathbf{e}(k) = \mathbf{y}(k) - \mathbf{y}_d(k)$$

$$\Delta \mathbf{X}(k+1) = \mathbf{A} \Delta \mathbf{X}(k) + \mathbf{B} \Delta \mathbf{u}(k)$$

$$\mathbf{e}(k+1) = \mathbf{e}(k) + \mathbf{C} \mathbf{A} \Delta \mathbf{X}(k) + \mathbf{C} \mathbf{B} \Delta \mathbf{u}(k) - \Delta \mathbf{y}_d(k+1)$$

## Preview Control

$$\bar{\mathbf{X}}(k) = [\mathbf{e}^T(k), \Delta \mathbf{X}^T(k), \Delta \mathbf{y}_d^T(k+1), \dots, \Delta \mathbf{y}_d^T(k+N)]^T \in \mathbb{R}^{p+n+pN}$$

$$\bar{\mathbf{X}}(k+1) = \bar{\mathbf{A}} \bar{\mathbf{X}}(k) + \bar{\mathbf{B}} \Delta \mathbf{u}(k)$$

$$\bar{\mathbf{A}} = \left[ \begin{array}{cc|cccc} \mathbf{I}_p & \mathbf{C}\mathbf{A} & -\mathbf{I}_p & 0 & \dots & \dots & 0 \\ 0 & \mathbf{A} & 0 & 0 & \dots & \dots & 0 \\ \hline 0 & 0 & 0 & \mathbf{I}_p & 0 & \dots & 0 \\ 0 & 0 & 0 & 0 & \mathbf{I}_p & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & \mathbf{I}_p \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right] \quad \bar{\mathbf{B}} = \left[ \begin{array}{c} \mathbf{C}\mathbf{B} \\ \mathbf{B} \\ 0 \\ \vdots \\ \vdots \\ \vdots \\ 0 \end{array} \right]$$

## Linear Optimal Control

$$J = \sum_{i=k}^{\infty} [\mathbf{e}^T(i) \mathbf{Q}_e \mathbf{e}(i) + \Delta \mathbf{X}^T(i) \mathbf{Q}_x \Delta \mathbf{X}(i) + \Delta \mathbf{u}^T(i) \mathbf{R} \Delta \mathbf{u}(i)] = \sum_{i=k}^{\infty} \bar{\mathbf{X}}^T(i) \bar{\mathbf{Q}} \bar{\mathbf{X}}(i) + \Delta \mathbf{u}^T(i) \bar{\mathbf{R}} \Delta \mathbf{u}(i)$$

$$\bar{\mathbf{Q}} = \left[ \begin{array}{c|c|c} \mathbf{Q}_e & \mathbf{0} & \mathbf{0} \\ \hline \mathbf{0} & \mathbf{Q}_x & \mathbf{0} \\ \hline \mathbf{0} & \mathbf{0} & \mathbf{0}_{pN} \end{array} \right]$$

$$\Delta \mathbf{u}^*(k) = -\mathbf{G}_1 \mathbf{e}(k) - \mathbf{G}_2 \Delta \mathbf{X}(k) - \sum_{i=1}^N \mathbf{G}_3(i) \Delta \mathbf{y}_d(k+i)$$

$$\mathbf{G}_1 = \mathbf{W} \mathbf{P} \tilde{\mathbf{I}} \quad \mathbf{G}_2 = \mathbf{W} \mathbf{P} \tilde{\mathbf{F}} \quad \mathbf{G}_3 = -\mathbf{W} \left( [\tilde{\mathbf{A}} - \tilde{\mathbf{B}} \mathbf{W} \mathbf{P} \tilde{\mathbf{A}}]^T \right)^{k-1} \mathbf{P} \tilde{\mathbf{I}}$$

$$\tilde{\mathbf{B}} = \begin{bmatrix} \mathbf{C} \mathbf{B} \\ \mathbf{B} \end{bmatrix} \quad \tilde{\mathbf{I}} = \begin{bmatrix} \mathbf{I}_p \\ \mathbf{0}_p \end{bmatrix} \quad \tilde{\mathbf{F}} = \begin{bmatrix} \mathbf{C} \mathbf{A} \\ \mathbf{A} \end{bmatrix} \quad \tilde{\mathbf{Q}} = \begin{bmatrix} \mathbf{Q}_e & \mathbf{0}_{p \times n} \\ \mathbf{0}_{n \times p} & \mathbf{Q}_x \end{bmatrix}$$

$$\tilde{\mathbf{A}} = [\tilde{\mathbf{I}} \quad \tilde{\mathbf{F}}] \quad \mathbf{W} = [\mathbf{R} + \tilde{\mathbf{B}}^T \mathbf{P} \tilde{\mathbf{B}}]^{-1} \tilde{\mathbf{B}}^T$$

## Linear Optimal Control

**Discrete Algebraic Riccati Equation (DARE)**

$$\mathbf{P} = \tilde{\mathbf{A}}^T \mathbf{P} \tilde{\mathbf{A}} - \tilde{\mathbf{A}} \mathbf{P} \tilde{\mathbf{B}} [\mathbf{R} + \tilde{\mathbf{B}}^T \mathbf{P} \tilde{\mathbf{B}}]^{-1} \tilde{\mathbf{B}}^T \mathbf{P} \tilde{\mathbf{A}} + \tilde{\mathbf{Q}}$$

$$\mathbf{u}^{optimal}(0) = -\mathbf{G}_2 \mathbf{X}(0) - \sum_{j=1}^N \mathbf{G}_3(j) \mathbf{y}_d(j)$$

$$\mathbf{u}^{optimal}(k) = -\mathbf{G}_1 \sum_{i=0}^k \mathbf{e}(i) - \mathbf{G}_2 \mathbf{X}(k) - \sum_{j=1}^N \mathbf{G}_3(j) \mathbf{y}_d(k+j)$$

**Error integral      State F/B      F/F(preview)**



## ZMP-Preview Biped Walking

