

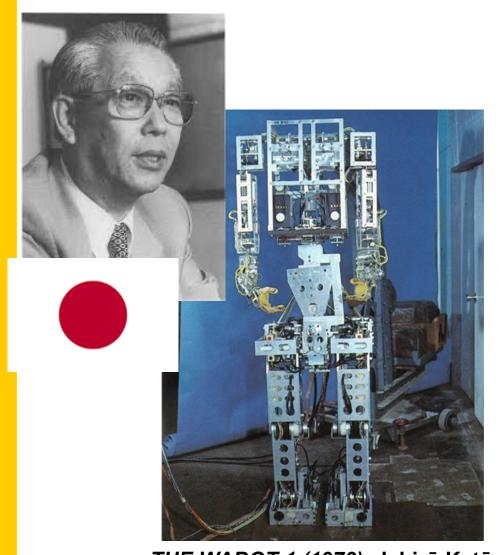


Walking Gaits for the NUgus Platform

... from MCHA 4100 to idealised walking strategies

Darcy Byrne C3256634

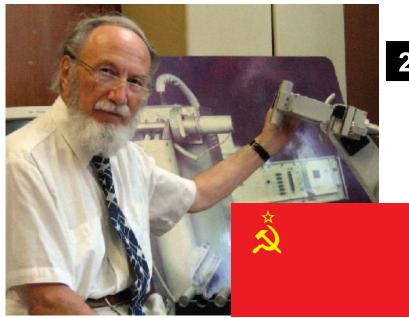
History



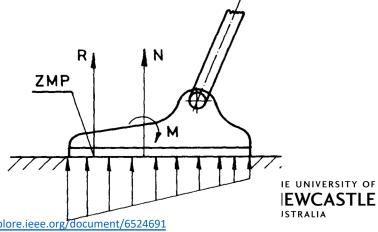
THE WABOT-1 (1973), Ichirō Katō "Quasi-Static" Locomotion

https://link.springer.com/article/10.1007/BF00735435?noAccess=true

https://www.humanoid.waseda.ac.jp/booklet/kato 2.html https://ieeexplore.ieee.org/document/6524691



Mathematical Models of General Anthropomorphic Systems (1973), Miomir Vukobratović "ZMP" Locomotion



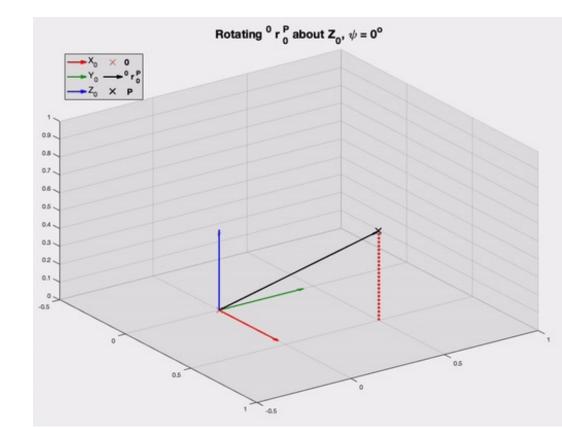
Mathematical Background

- Vectors
 Direction & Magnitude, with respect to a Coordinate System
- Rotations (Passive & Active)
 Coordinate transformers

$$\mathbf{r}_P^0 = \begin{bmatrix} 0.5 & 0.5 & 0.5 \end{bmatrix}^T$$

$$\mathbf{R}_{\mathbf{Z}_0}(\psi) = \begin{bmatrix} \cos \psi & -\sin \psi & 0 \\ \sin \psi & \cos \psi & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{R}_{\mathbf{Z}_0}^0 \mathbf{r}_P^0 \implies$$



Canonical Homogenous Transforms

$$A_{\mathbf{b}}^{\mathbf{a}}(q) = \begin{bmatrix} R_{\mathbf{b}}^{\mathbf{a}}(q) & r_{\mathbf{b}}^{\mathbf{a}}(q) \\ \mathbf{0} & 1 \end{bmatrix} = A_{\mathbf{a}}^{\mathbf{b}}(q)^{-1}$$

Kinematic Chains

$$T_n^0(q) = A_1^0(q_1)A_2^1(q_2) \dots A_n^{n-1}(q_n) = \begin{bmatrix} R_n^0(q) & r_n^0(q) \\ 0 & 1 \end{bmatrix}$$

... where q is an array of 'Joint Variables'

$$T_n^0(q) \Rightarrow$$
 Forward Kinematic Model $\Rightarrow k(q)$



- Forward Kinematic Model (FKM)
 - Returns the End Effector Position, $r_{x_e}^0$, with respect to the **ZERO coordinate system**
 - Returns End Effector Orientation, $\Theta_{x_e}^0$, with respect to the **ZERO coordinate system**
 - However, this first involves the parameterisation of the Rotation Matrix $R_{x_0}^0(q)$ into Euler Angles or Unit Quaternions
 - 'Google it, mate.'

- Comrade Adam Bandt, 2022

$$x_e = \begin{bmatrix} r_{x_e}^0 \\ \boldsymbol{\Theta}_{x_e}^0 \end{bmatrix} = k(q) = [x, y, z, \varphi, \theta, \psi]^T$$



- Inverse Kinematics (IKM)
 - Returns the array of Joint Variables necessary to achieve a desired End Effector Pose, x_e
 - Usually solved with Numerical Optimisation
- Parallel Kinematics



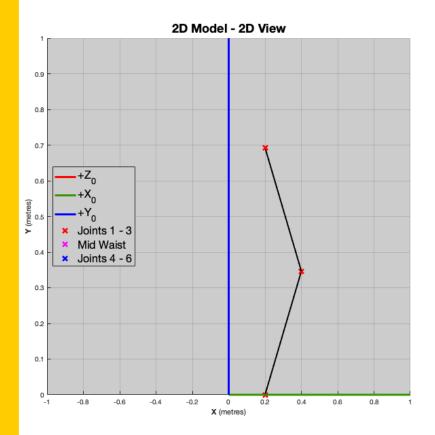
Numerical Optimisation
 General Solution to the IKM problem:

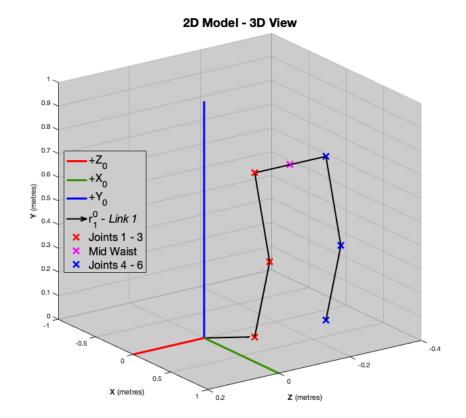
$$q^* = \arg\min_{q} K_q ||q_0 - q|| + K_e ||k(q) - x_e^*||$$

... which is to say; the optimal array of **State Variables**, q^* , that satisfy the desired **End Effector Pose**, x_e^* , can be found by minimising the above function.



2D Model





$$\boldsymbol{T}_6^1 = \boldsymbol{A}_2^1 \boldsymbol{A}_3^2 \boldsymbol{A}_4^3 \boldsymbol{A}_5^4 \boldsymbol{A}_6^5$$

$$T_1^6 = (T_6^1)^{-1}$$

$$T_1^{Global} = A_1^0 T_6^1$$

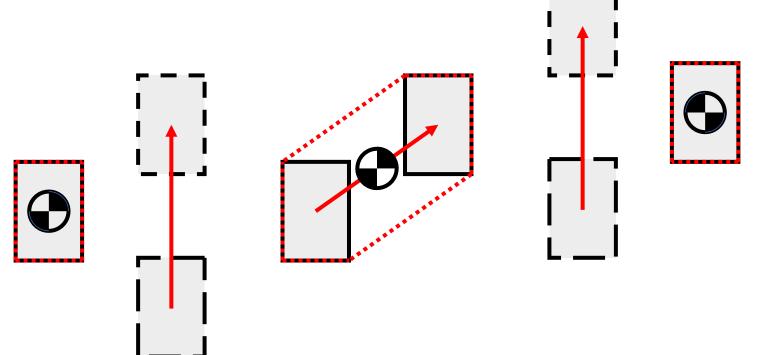
 $T_6^{Global} = A_6^0 T_1^6$

$$q = [\psi_1, \psi_2, \psi_3, \psi_4, \psi_5, \psi_6]^T$$



Quasi Static

$$r_{CoM}^0 = \frac{1}{\sum Mass} \sum_{i=1}^n m_i r_i^0$$



"Support Polygon"



Quasi Static – Cont'd...

Single Support *IKM*:

$$\mathbf{q}^* = \underset{\mathbf{q}}{\operatorname{arg \, min}} \ \mathbf{K}_q \|\mathbf{q_0} - \mathbf{q}\| + \mathbf{K}_e \|k(\mathbf{q}) - \mathbf{x}_e\| + \mathbf{K}_{CoM} \|\mathbf{r}_{CoM}^0 - \mathbf{r}_{Fixed}^0\|$$

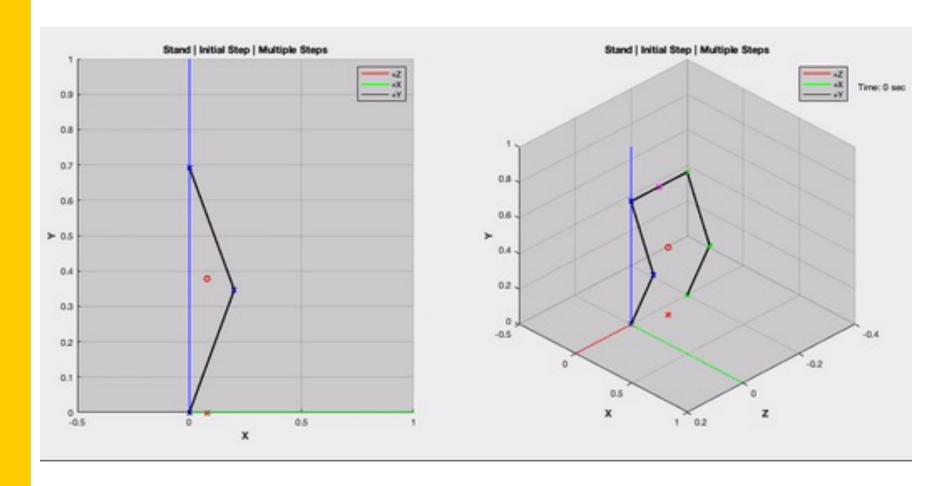
Double Support *IKM*:

$$\mathbf{q}^* = \underset{\mathbf{q}}{\operatorname{arg \, min}} \ \mathbf{K}_q \|\mathbf{q_0} - \mathbf{q}\| + \mathbf{K}_e \|k(\mathbf{q})\| + \mathbf{K}_{CoM} \|\mathbf{r}_{CoM}^0 - \mathbf{r}_{Forward \, foot}^0\|$$

- 1. Single Support
 - 1. Generate Step Trajectory
 - 2. Recursively Solve **IKM** for each point along trajectory
- 2. Double Support
 - 1. Generate **CoM** Trajectory
 - 2. Recursively Solve **IKM** for each point along trajectory

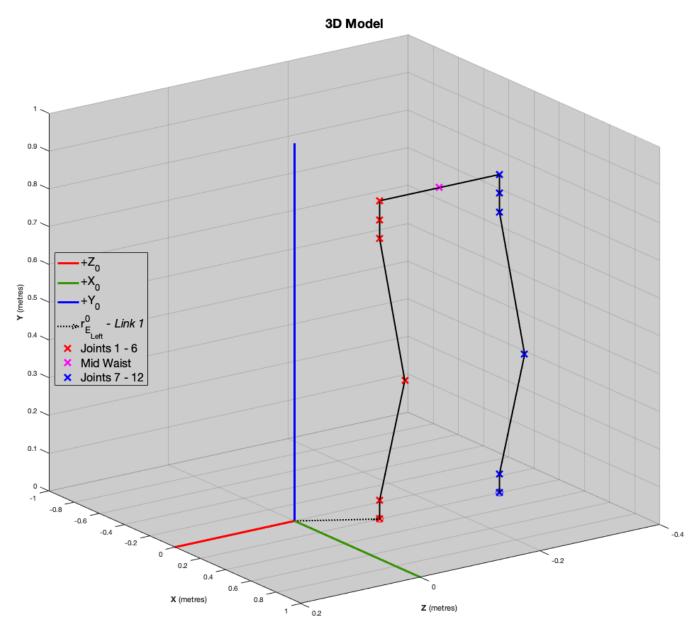


2D Model Walking



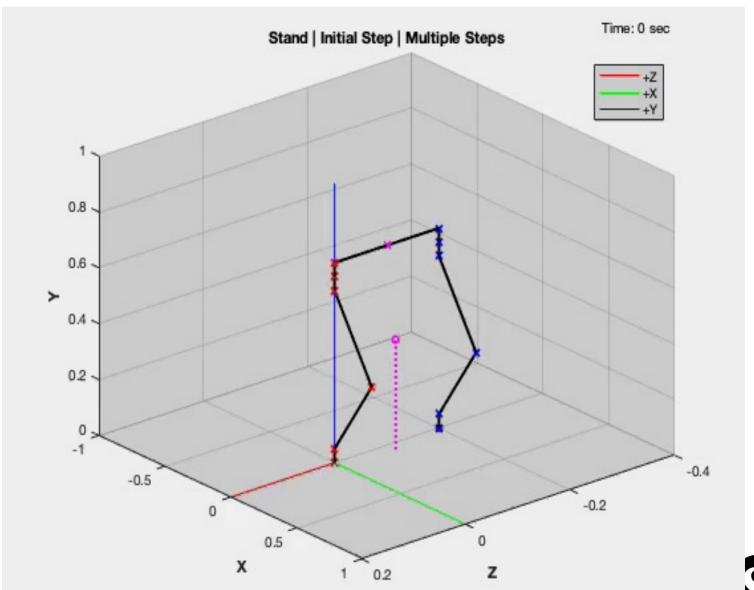


3D Model





3D Model Walking – Quasi Static





Zero Moment Point

Supposed, while standing on one leg, that **YOU** are an **inverted pendulum** ...

Your entire degree has led to this moment ...
You begin to fall ...

But at the last moment you place your foot down, shifting all your mass to it ...

You're now walking using **ZMP**



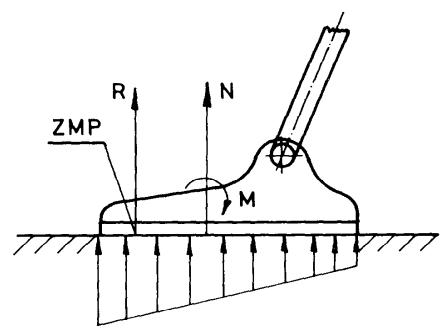


Fig. 3. Load distribution by foot longitudinal section.

Mathematical Model of General Anthropomorphic Systems

- Vukobratović, et al. 1973

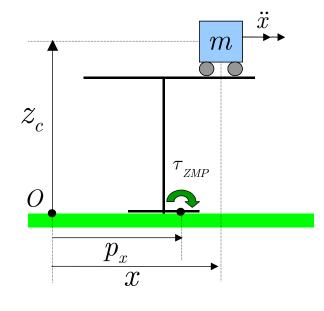
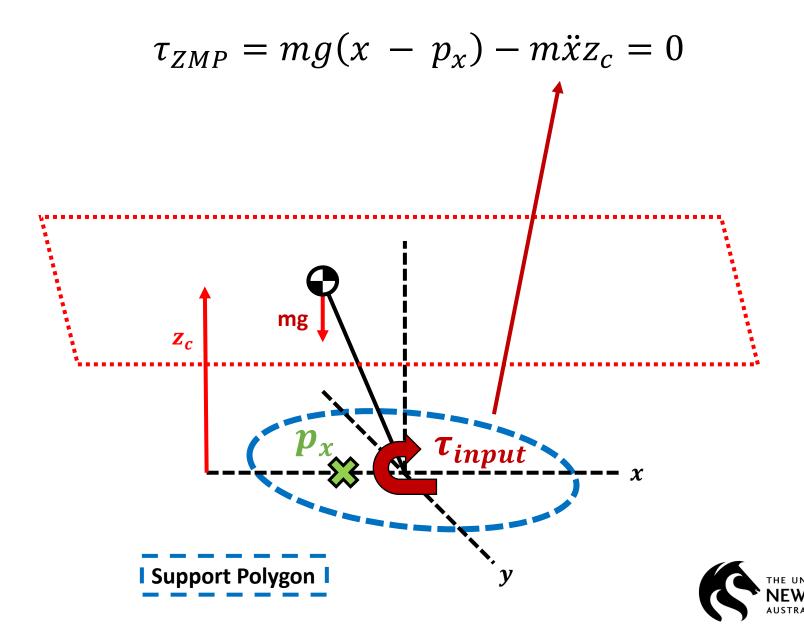


Figure 3: A cart-table model

Biped Walking Pattern Generation by using Preview Control of Zero-Moment Point

- Kajita, et al. 2003





$$\frac{\delta}{\delta t} \begin{bmatrix} x \\ \dot{x} \\ \dot{x} \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \\ \dot{x} \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} u_{(x,y)}$$

$$\mathbf{p}_{(x,y)} = \begin{bmatrix} 1 & 0 & \frac{-z_c}{g} \end{bmatrix} \begin{bmatrix} x \\ \dot{x} \\ \ddot{x} \end{bmatrix}$$

$$\mathbf{x}_{(k+1)} = \mathbf{A_d} \mathbf{x}_k + \mathbf{B_d} \mathbf{u}_k$$

 $\mathbf{p}_k = \mathbf{C_d} \mathbf{x}_k$

... where:

$$\mathbf{A_d} = \begin{bmatrix} 1 & T & \frac{T^2}{2} \\ 0 & 1 & T \\ 0 & 0 & 1 \end{bmatrix} \ \mathbf{B_d} = \begin{bmatrix} \frac{T^3}{6} \\ \frac{T^2}{2} \\ T \end{bmatrix} \ \mathbf{C_d} = \begin{bmatrix} 1 & 0 & \frac{-z_c}{g} \end{bmatrix}$$



$$J_{u} = \sum_{i=k}^{\infty} \mathbf{Q}_{e} \|\mathbf{y}(k) - \mathbf{y}_{d}(k)\| + \mathbf{Q}_{x} \|\mathbf{x}(k) - \mathbf{x}(k-1)\| + \mathbf{R} \|\mathbf{u}(k) - \mathbf{u}(k-1)\|$$

$$\mathbf{u}^*(k) = -\mathbf{G_e} \sum_{i=0}^{k} [\mathbf{y}(k) - \mathbf{y_d}(k)] - \mathbf{G_x} \mathbf{x}(k) - \sum_{l=1}^{N_l} \mathbf{G_d}(l) \mathbf{y_d}(k+l)$$

$$\tilde{\mathbf{B}} = \begin{bmatrix} \mathbf{C_d} \mathbf{B_d} \\ \mathbf{B_d} \end{bmatrix}, \ \ \tilde{\mathbf{F}} = \begin{bmatrix} \mathbf{C_d} \mathbf{A_d} \\ \mathbf{A_d} \end{bmatrix}, \ \ \tilde{\mathbf{Q}} = \begin{bmatrix} \mathbf{Q_e} & \mathbf{0} \\ \mathbf{0} & \mathbf{Q_x} \end{bmatrix}, \ \ \tilde{\mathbf{I}} = \begin{bmatrix} \mathbf{I} \\ \mathbf{0} \end{bmatrix}, \ \ \tilde{\mathbf{A}} = \begin{bmatrix} \tilde{\mathbf{I}} & \tilde{\mathbf{F}} \end{bmatrix}$$

... where the gains, G_e , G_x and $G_d(l)$ are given by;

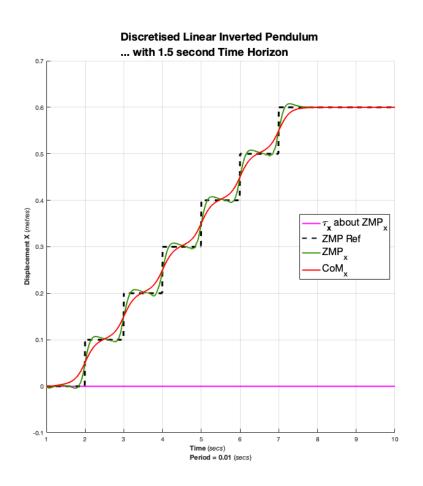
$$\begin{split} \mathbf{G_e} &= [\mathbf{R} + \tilde{\mathbf{B}}^\intercal \tilde{\mathbf{K}}_\mathbf{d} \tilde{\mathbf{B}}]^{-1} \tilde{\mathbf{B}}^\intercal \tilde{\mathbf{K}}_\mathbf{d} \tilde{\mathbf{I}} \\ \mathbf{G_x} &= [\mathbf{R} + \tilde{\mathbf{B}}^\intercal \tilde{\mathbf{K}}_\mathbf{d} \tilde{\mathbf{B}}]^{-1} \tilde{\mathbf{B}}^\intercal \tilde{\mathbf{K}}_\mathbf{d} \tilde{\mathbf{F}} \\ \mathbf{G_d}(l) &= -[\mathbf{R} + \tilde{\mathbf{B}}^\intercal \tilde{\mathbf{K}}_\mathbf{d} \tilde{\mathbf{B}}]^{-1} \tilde{\mathbf{B}}^\intercal (\tilde{\mathbf{A}}_\mathbf{c}^\intercal)^{(l-1)} \tilde{\mathbf{K}}_\mathbf{d} \tilde{\mathbf{I}} \end{split}$$

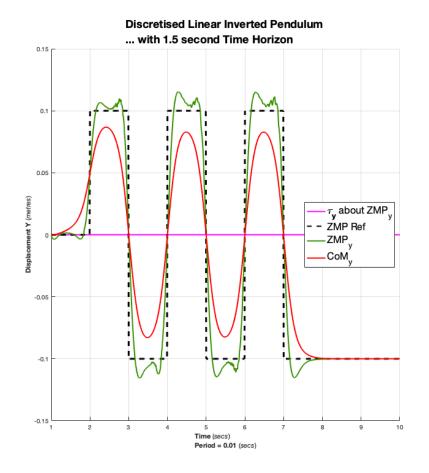
... where;

$$\tilde{\mathbf{A}}_{\mathbf{c}} = \tilde{\mathbf{A}} - \tilde{\mathbf{B}}[\mathbf{R} + \tilde{\mathbf{B}}^{\intercal} \tilde{\mathbf{K}}_{\mathbf{d}} \tilde{\mathbf{B}}]^{-1} \tilde{\mathbf{B}}^{\intercal} \tilde{\mathbf{K}}_{\mathbf{d}} \tilde{\mathbf{A}}$$

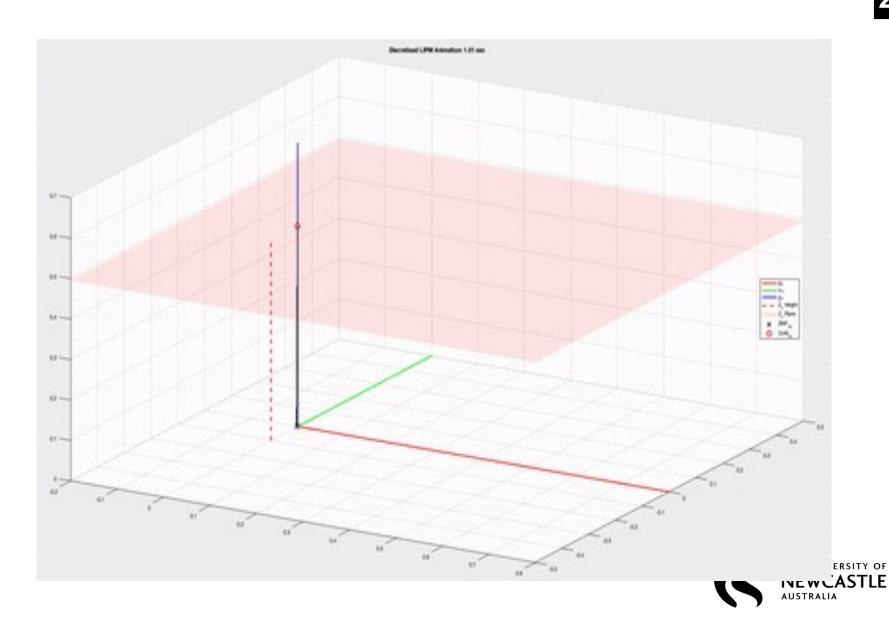
... and $\tilde{\mathbf{K}}_{\mathbf{d}}$ is a solution to the Discrete Time Algebraic Riccati Equation.



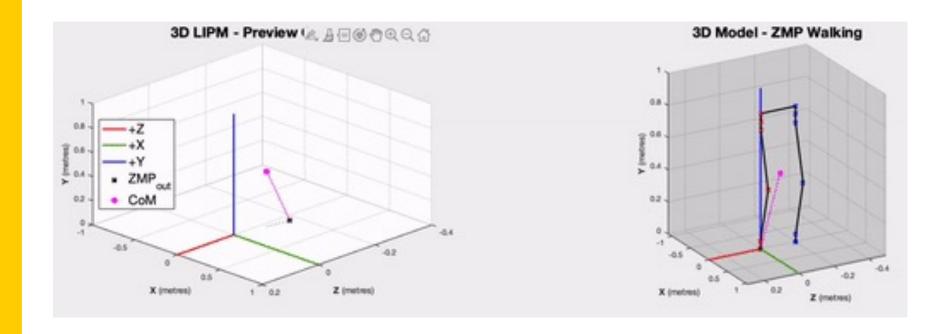








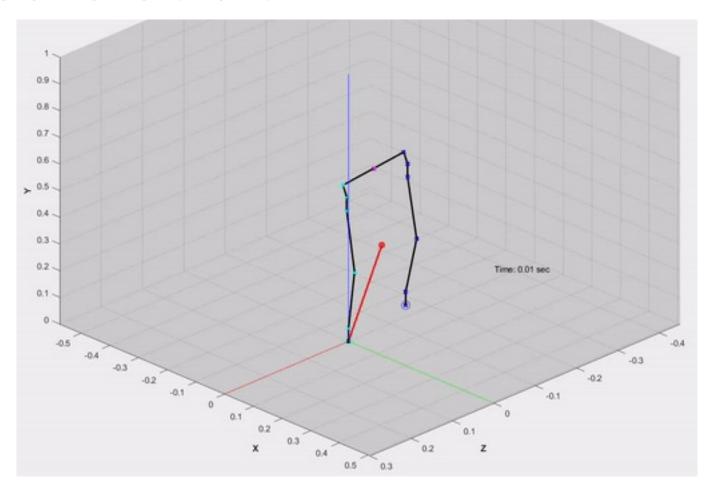
3D Model Walking – ZMP





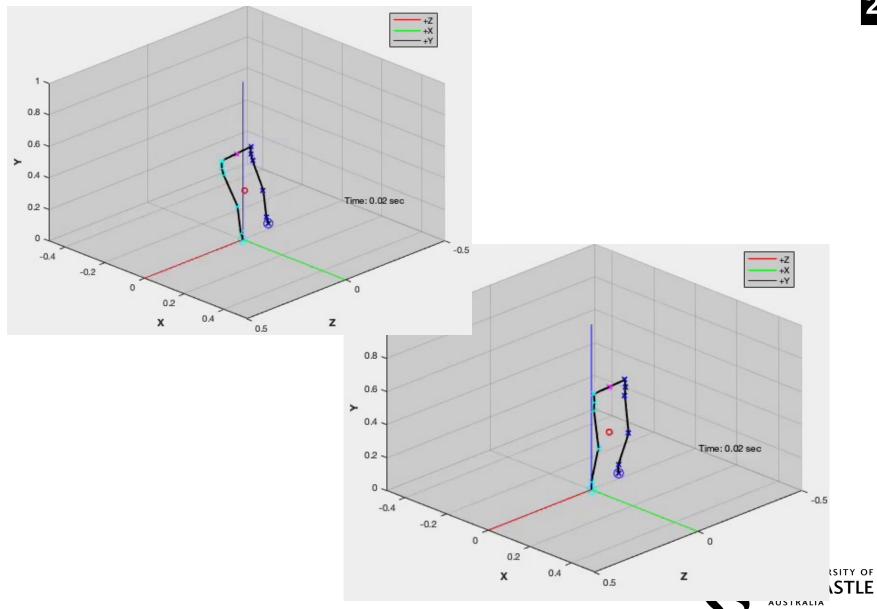
3D Model Walking – ZMP

Zero Moment Point





FAILS



Next Steps...

Simulation:

- 1. Shift from Global Coordinate System to Body Fixed
- 2. Optimal Trajectory for each foot step
- 3. Torso and Arms
- 4. Following a predefined path
- 5. Disturbance Rejection
- 6. Spring Loaded Invert Pendulum

Reality:

- 1. MatLab Plugin: `Realtime` Joint Variables to WeBots
- 2. Quasi Static Locomotion implemented on NUgus
- Zero Moment Point Locomotion implemented on NUgus

