

# Humanoid Robotics

## Bipedal Walking

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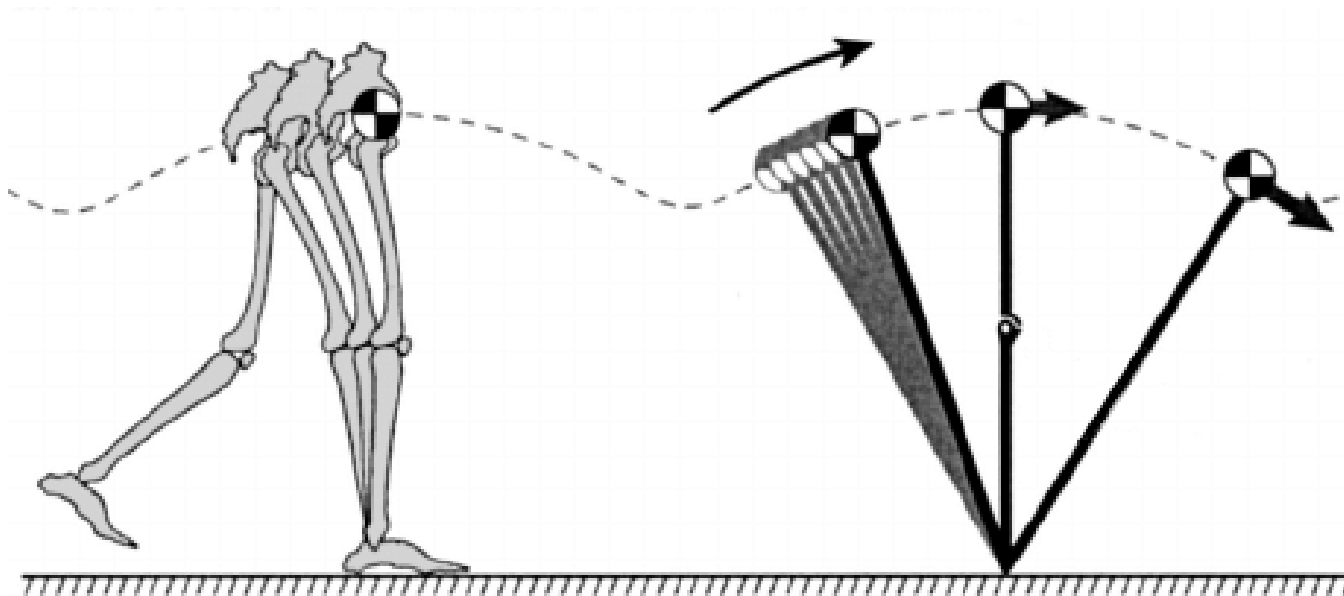


# Motivation

- Bipedal walking is the most versatile form of locomotion
- Legs can go anywhere: walk, jump, run, climb, and swim
- Legs can also be used to drive cars and to ride bicycles

# Motivation

- Walking on stretched legs is energy-efficient



Source: Omar et.al.: Study of Bipedal Robot Walking Motion in Low Gravity: Investigation and Analysis

# Motivation

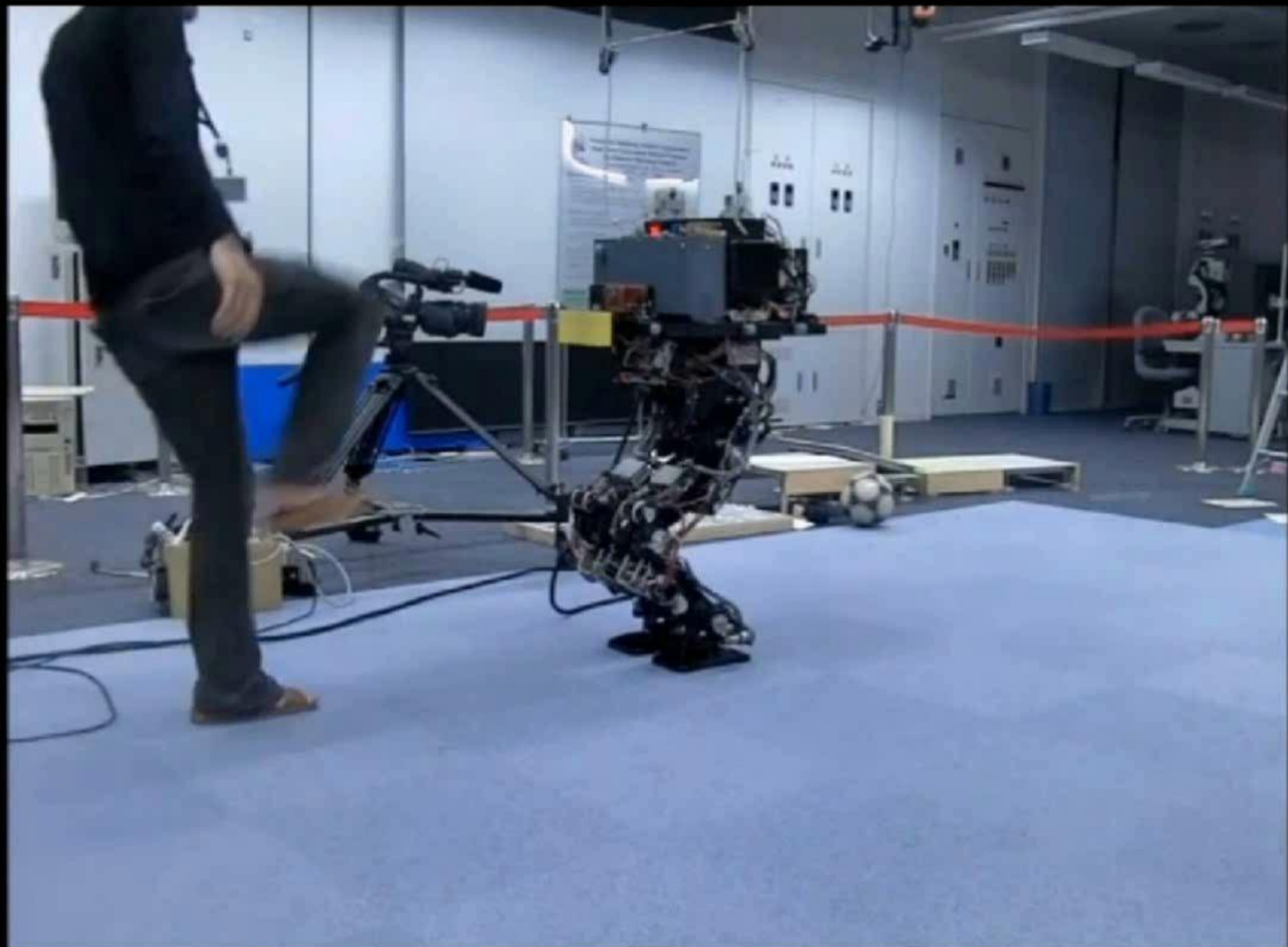
- Bipedal walking is the most versatile form of locomotion
- Legs can go anywhere: walk, jump, run, climb, and swim
- Walking on stretched legs is energy-efficient
- However: **balancing is difficult**















The robot attempts to maintain the reference velocity after a disturbance.

# Passive Dynamic Walking

- Walking only with a mechanical construction
- No control and no actuation apart from a shallow slope
- Proves that walking is strongly based on the natural dynamics of legs



Nagoya Inst. Tech. June 2005. About 4000 steps (about 35 minutes).  
Yoshito Ikemata, Akihito Sano & Hideo Fujimoto



# Central Pattern Generated Walking

- We add actuation and control and obtain open-loop walking robots
- Parameterized periodic motion patterns are used to generate targeted stepping motions
- Control of the walking velocity and direction
- Still **no control of balance!**





# Kinematics



# Kinematics

- Kinematics describes the geometry and the **motion of physical bodies.**
- Position, velocity, and acceleration are considered, but no forces (that would be dynamics).
- The motion of kinematic chains is described by **rigid body transformations.**

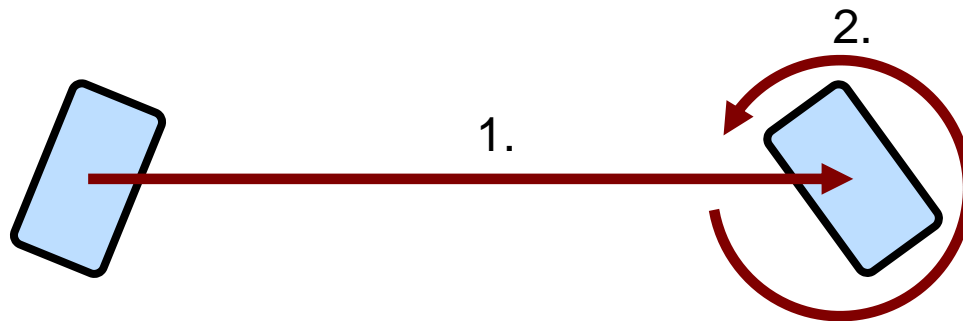
# Kinematics

- **Forward kinematics** computes the pose of a robot using the measured joint angles.
- **Inverse kinematics** computes the joint angles needed to reach a specific pose or end effector position.
- **Abstract kinematics** computes joint angles from abstract leg pose parameters. This has certain advantages over inverse kinematics.

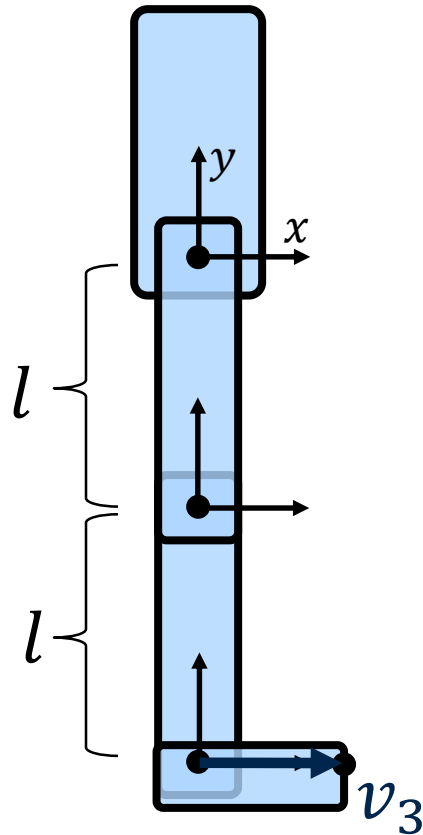
# Rigid Body Transform

- A rigid body transformation consisting of a translation by a vector  $(d_x, d_y)$  and then a rotation by an angle  $\alpha$  can be represented by a multiplication with a homogeneous transformation matrix.

$$\begin{bmatrix} v_x' \\ v_y' \\ 1 \end{bmatrix} = \begin{bmatrix} \cos \alpha & -\sin \alpha & d_x \\ \sin \alpha & \cos \alpha & d_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} v_x \\ v_y \\ 1 \end{bmatrix}$$



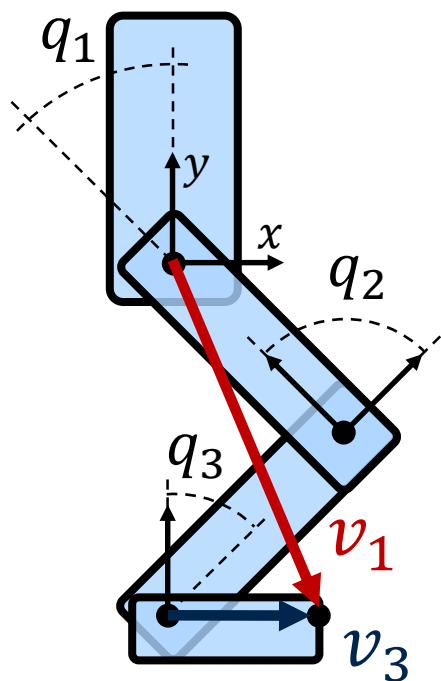
# Forward Kinematics



- We begin with a known **kinematic model**, i.e., the lengths of the links of the kinematic chain and the position of the end effector  $v_3$ .
- The **zero pose** is a special pose calibrated such that the transformation between the links is known and the joint angles  $q_1$ ,  $q_2$ , and  $q_3$  are zero.

# Forward Kinematics (2)

- By chaining multiple rigid body transforms, we can compute the position of the end effector in the base frame.

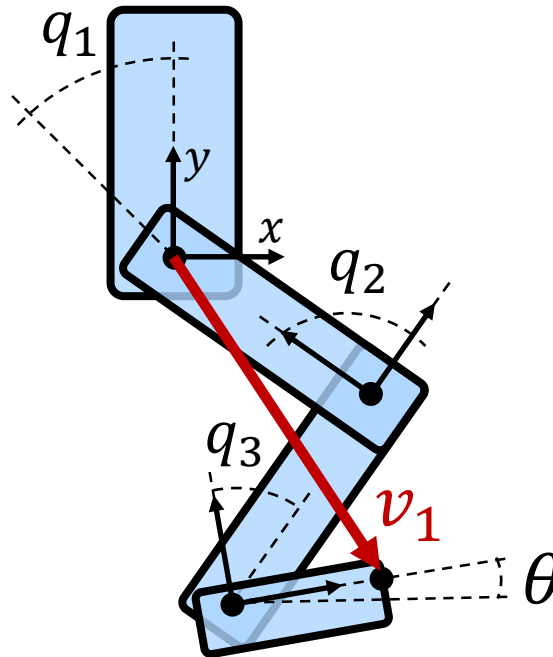


$$T_1(q_1) = \begin{bmatrix} \cos q_1 & -\sin q_1 & l \sin q_1 \\ \sin q_1 & \cos q_1 & -l \cos q_1 \\ 0 & 0 & 1 \end{bmatrix}$$

$$v_1 = T_1(q_1) \cdot T_2(q_2) \cdot T_3(q_3) \cdot v_3$$

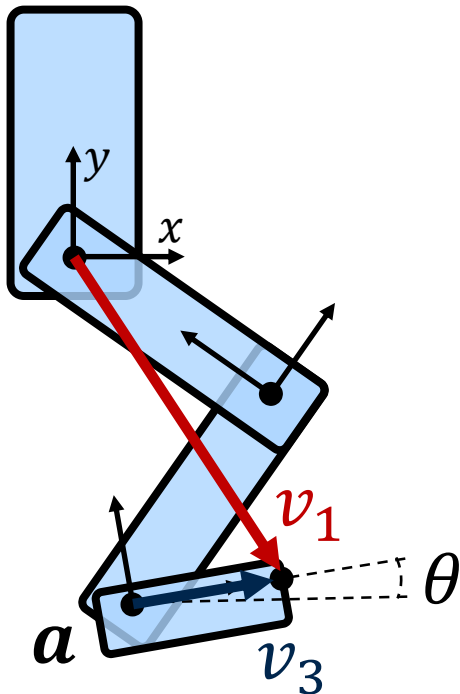
# Inverse Kinematics

- Inverse kinematics is the opposite of forward kinematics
- Input:  $v_1$  (end effector),  $\theta$  (foot angle)
- Output:  $q_1, q_2, q_3$  (joint angles)



# Inverse Kinematics (2)

- First, we compute the location of the ankle joint  $a$  using the end effector vectors  $v_1$  and  $v_3$  and the foot angle  $\theta$ .



$$a = (v_{1x} - |v_3| \cos \theta, v_{1y} - |v_3| \sin \theta)$$

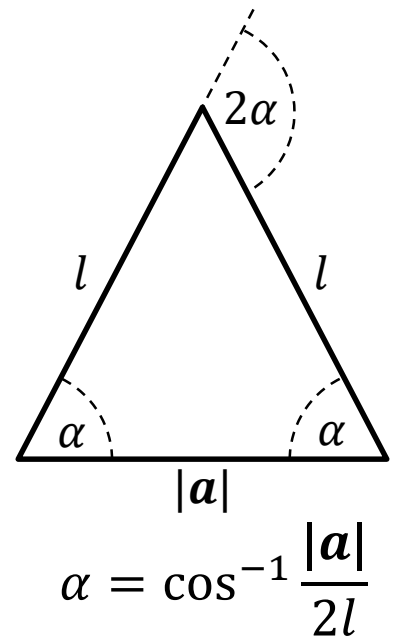
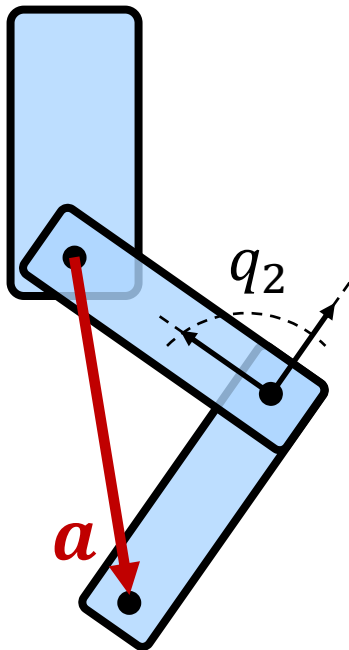


# Inverse Kinematics (3)

- Then, we compute the angle  $q_2$  of the knee joint using a triangle principle.

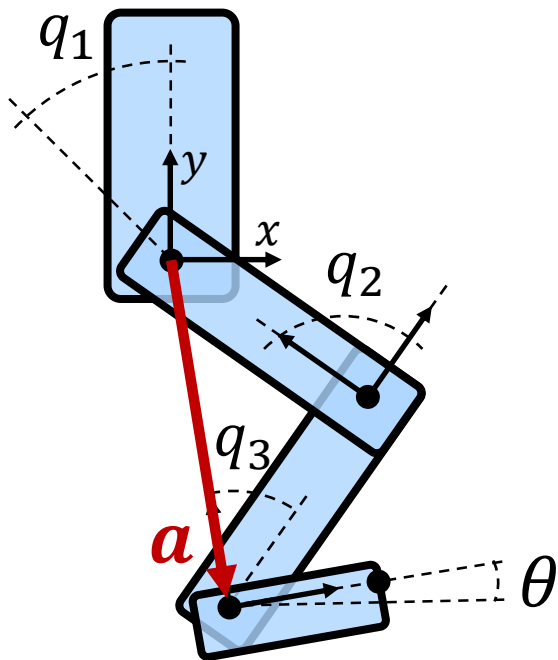
$$\mathbf{a} = (v_{1x} - |\mathbf{v}_3| \cos \theta, v_{1y} - |\mathbf{v}_3| \sin \theta)$$

$$q_2 = -2 \cos^{-1} \frac{|\mathbf{a}|}{2l}$$



# Inverse Kinematics (4)

- Now we can compute the remaining joint angles.

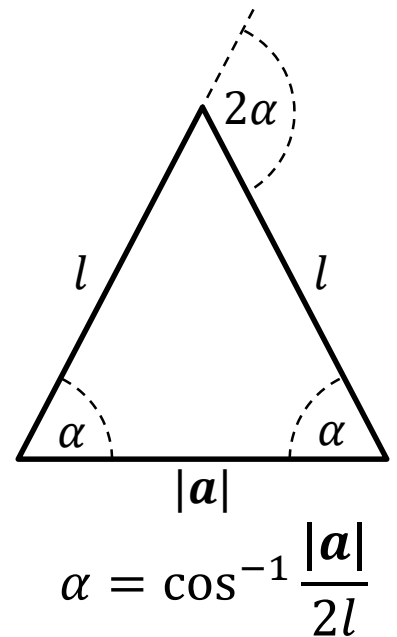


$$\mathbf{a} = (v_{1x} - |\mathbf{v}_3| \cos \theta, v_{1y} - |\mathbf{v}_3| \sin \theta)$$

$$q_2 = -2 \cos^{-1} \frac{|\mathbf{a}|}{2l}$$

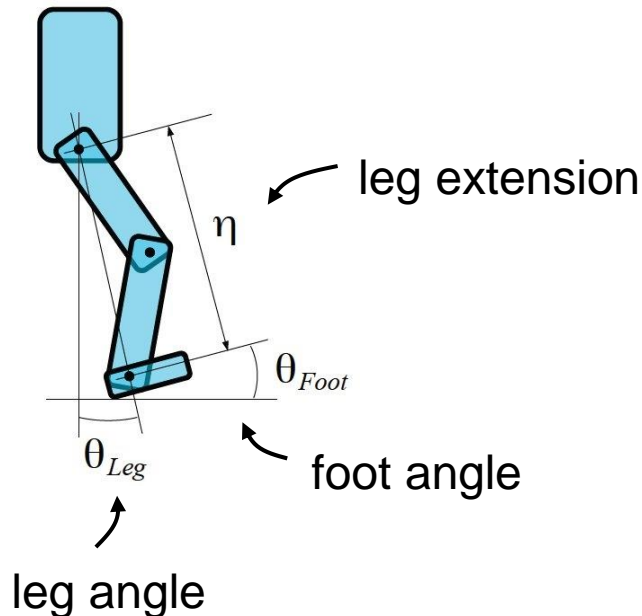
$$q_1 = \sin^{-1} \frac{a_x}{|\mathbf{a}|} - \frac{q_2}{2}$$

$$q_3 = \theta - q_1 - \frac{q_2}{2}$$



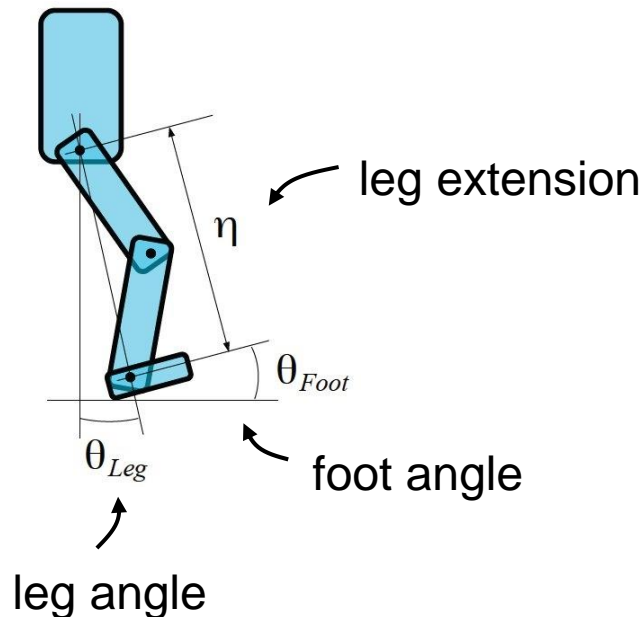
# Abstract Kinematics

- In Abstract Kinematics, the leg pose is parameterized in a different way as opposed to an end effector position. The computations are simpler and **no kinematic model is needed.**



# Abstract Kinematics (2)

- Input:  $\eta \in [0,1]$  (leg extension),  $\theta_{Leg}$  (leg angle),  $\theta_{Foot}$  (foot angle)
- Output:  $q_1, q_2, q_3$  (joint angles)



$$q_2 = 2 \cos^{-1} \eta$$

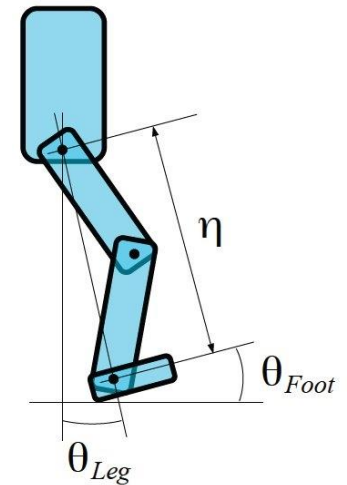
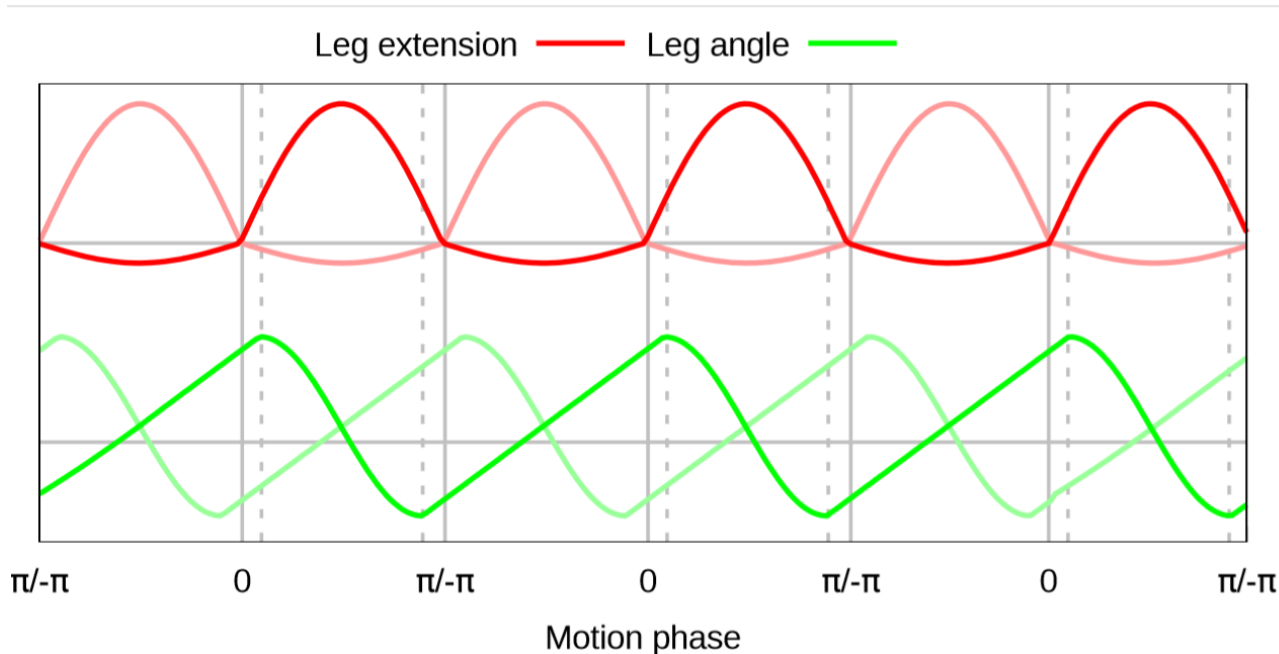
$$q_1 = \theta_{Leg} - \frac{q_2}{2}$$

$$q_3 = \theta_{Foot} - \theta_{Leg} - \frac{q_2}{2}$$

# Motion Pattern Generation

# Step Motion Pattern Generator

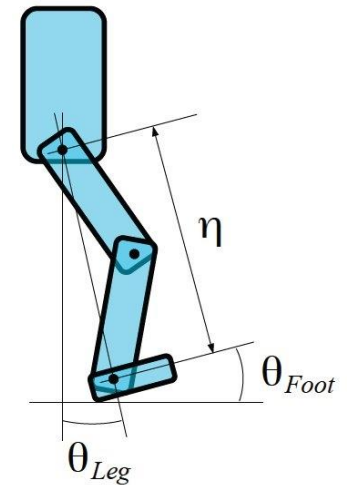
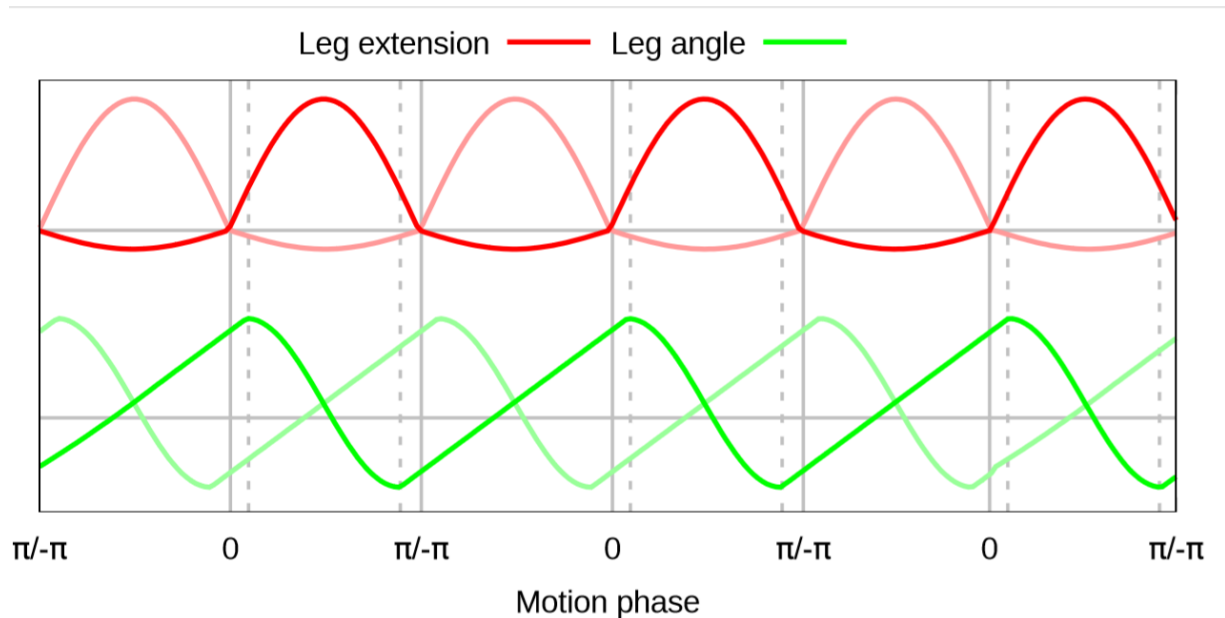
- Using either Inverse Kinematics or Abstract Kinematics, we can make a robot walk with the help of hand-crafted oscillating motion signals for the legs.



# Step Motion Pattern Generator

Leg Lifting: 
$$\eta(\mu) = \begin{cases} C_1 \sin \mu, & \mu < 0 \\ C_2 \sin \mu, & \mu \geq 0 \end{cases}$$

Leg Swing: 
$$\theta_{Leg}(\mu, S) = \begin{cases} S \left( \frac{2\mu + \pi}{\pi} \right), & \mu < 0 \\ S \cos \mu, & \mu \geq 0 \end{cases}$$





# Minimal Walking Function

```
while(true)
```

```
{
```

$$\mu = \mu + \Delta\mu$$

$$(q_1, q_2, q_3) = \text{abstractKinematics}(\eta(\mu), \theta_{Leg}(\mu, S))$$

```
sendToRobot( $q_1, q_2, q_3$ )
```

```
}
```

