

# Data Driven Computer Animation

**HKU COMP 7508**

**Tutorial 4: Real-time Character Control**

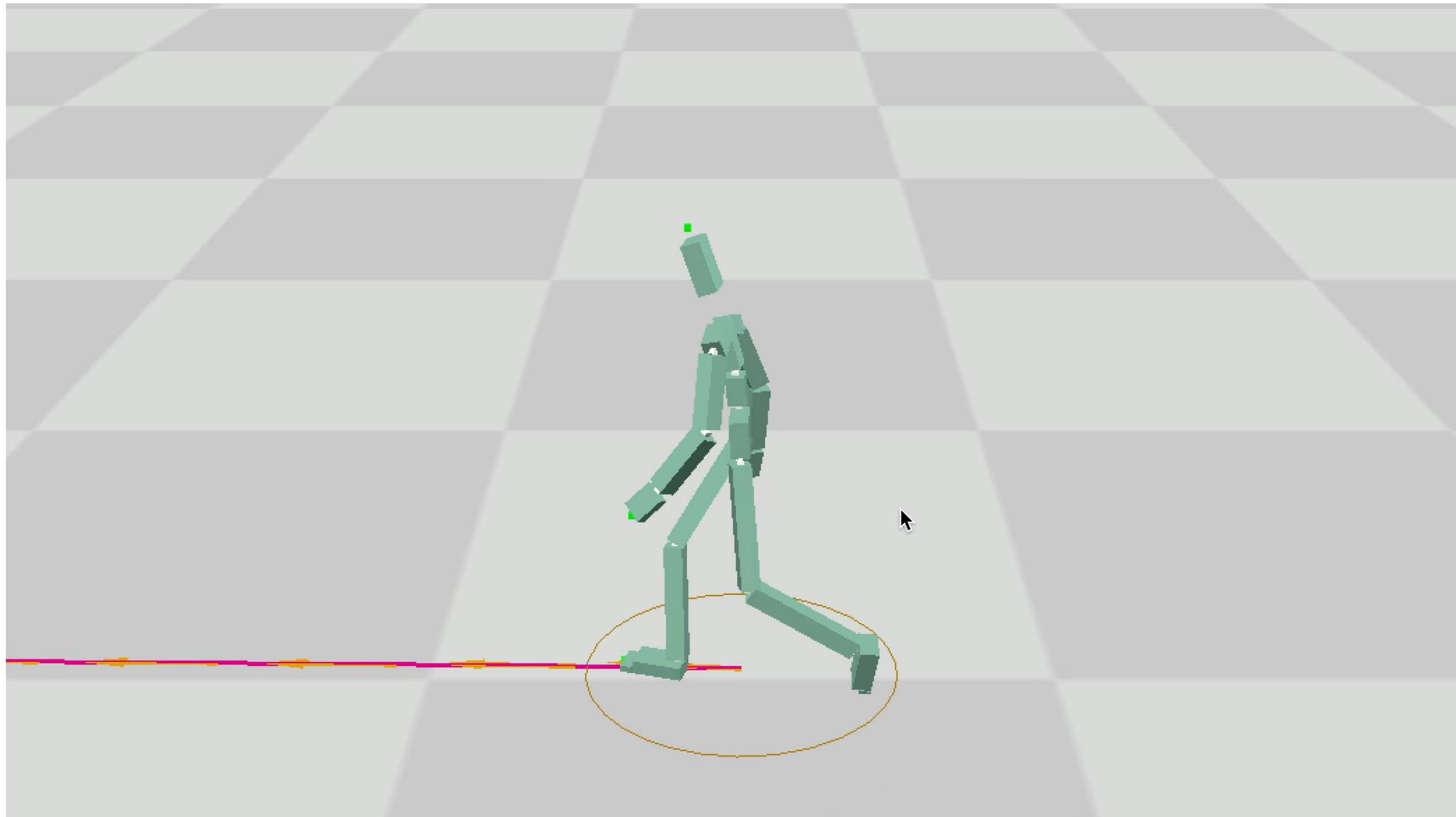
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**TA: Kemeng Huang ([kmhuang@connect.hku.hk](mailto:kmhuang@connect.hku.hk))**



# Agenda

1. The overview of real-time character control
2. State machine
3. Motion matching



Some materials are from:



# Game Developers Conference (GDC)

1. <https://www.gdcvault.com/play/1026968/-Genshin-Impact-Building-a>
2. <https://www.gdcvault.com/play/1025389/Character-Control-with-Neural-Networks>

# Animation

We captured motion clips, label them, and play them in different order





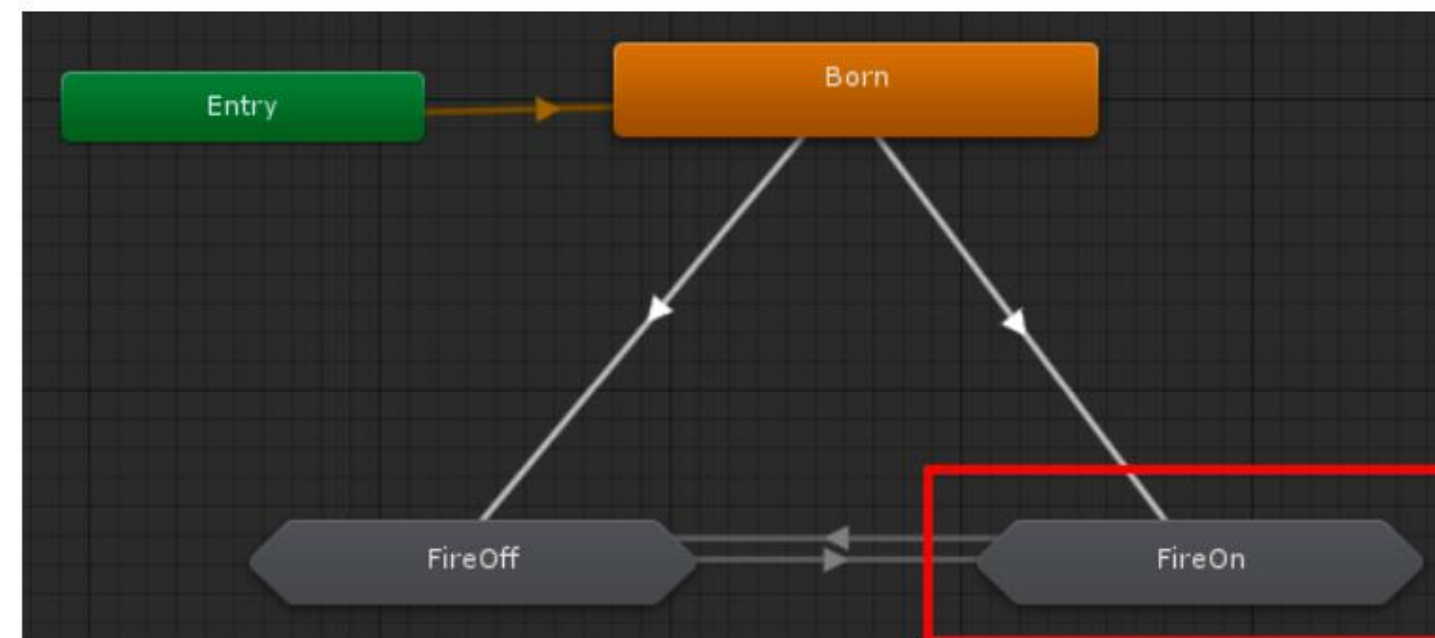
# How to make the animation to be controllable?

The character should reflect to different control signals

- from keyboard
- gamepad



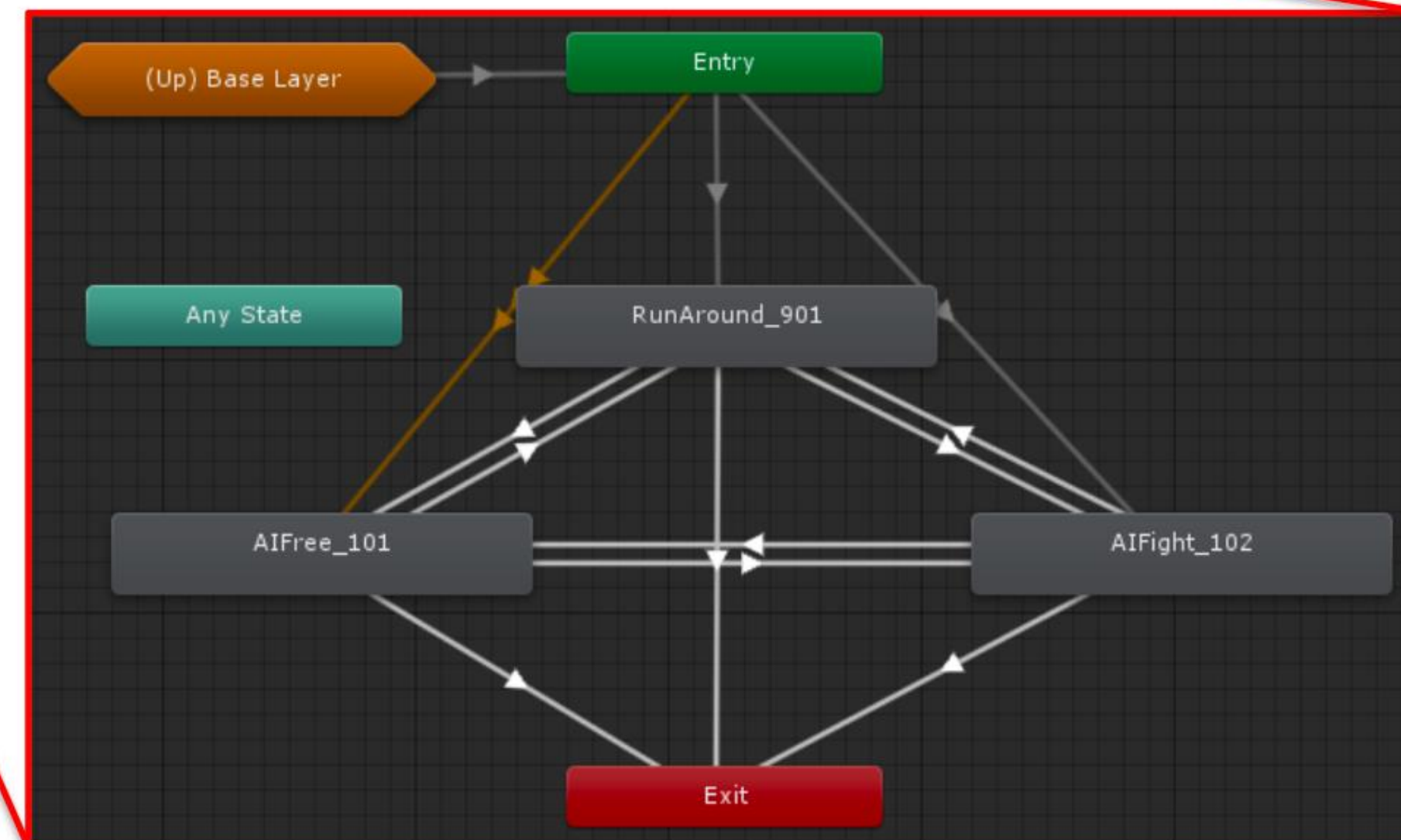
# The states are managed by 'node'



example of fire slime

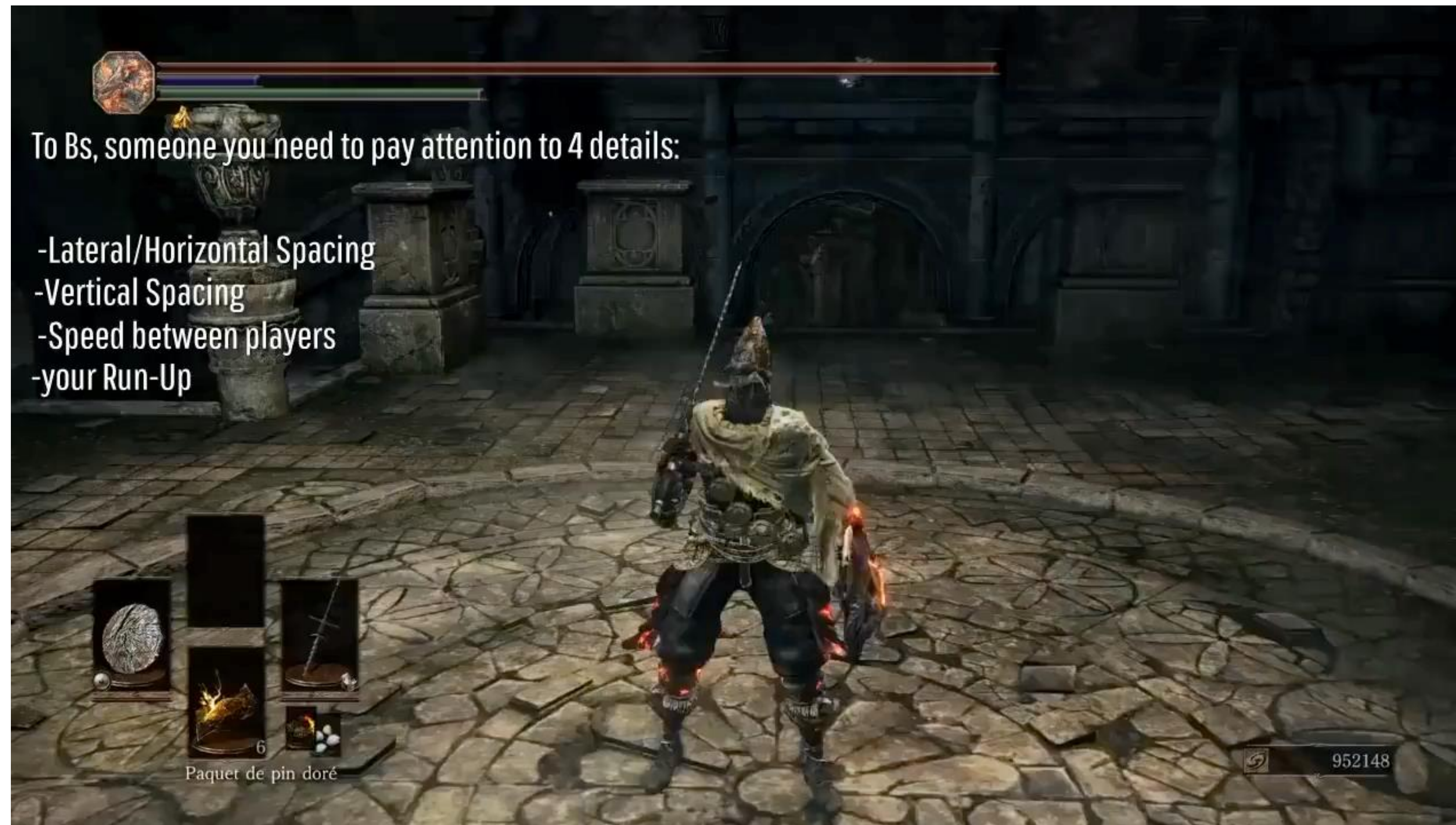


- A Layered System
- Node can be connected to any other node

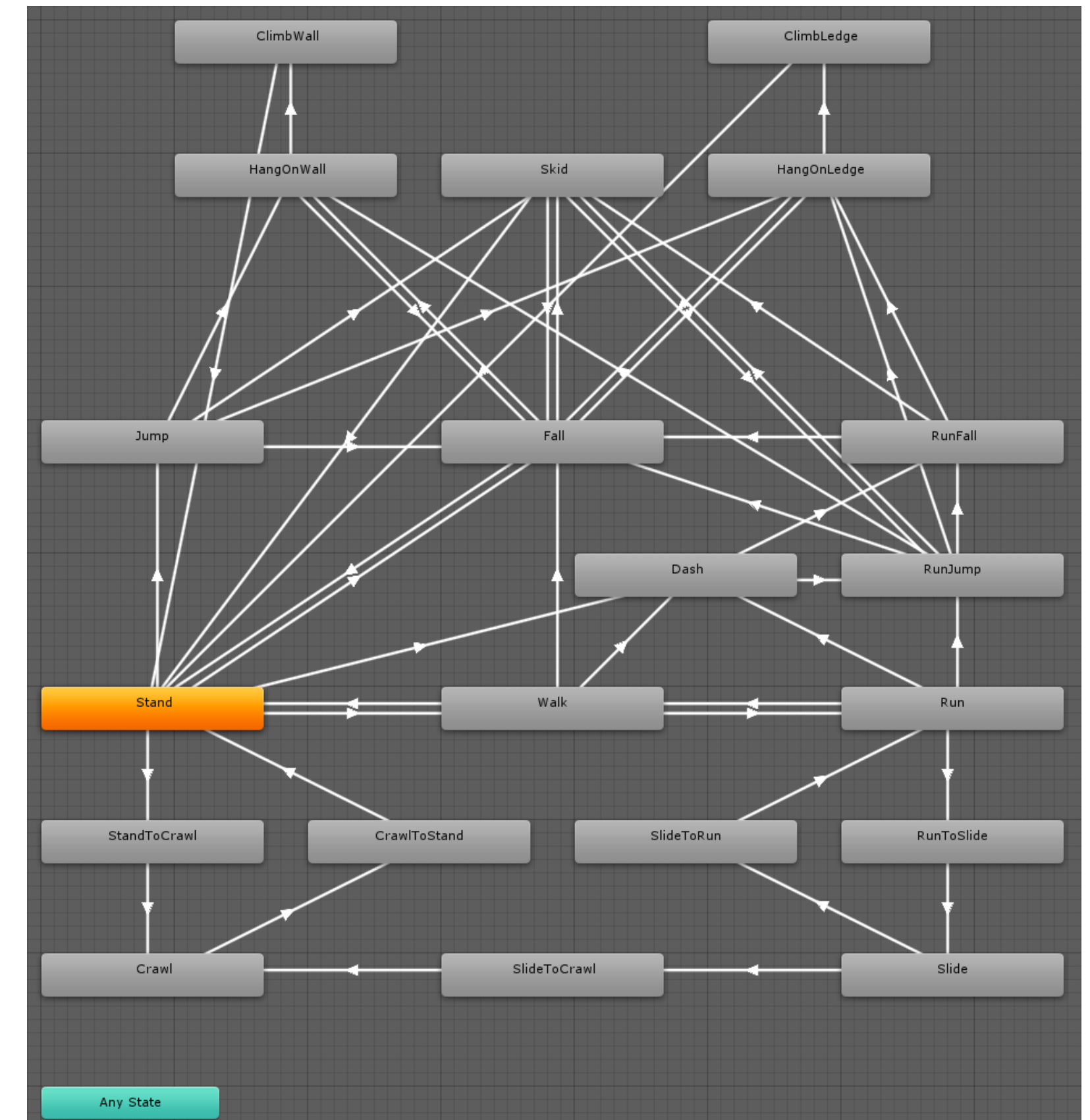




# Node system contains lots of transition



Controlling system is a combination for lots of motion clips



Character State Machine

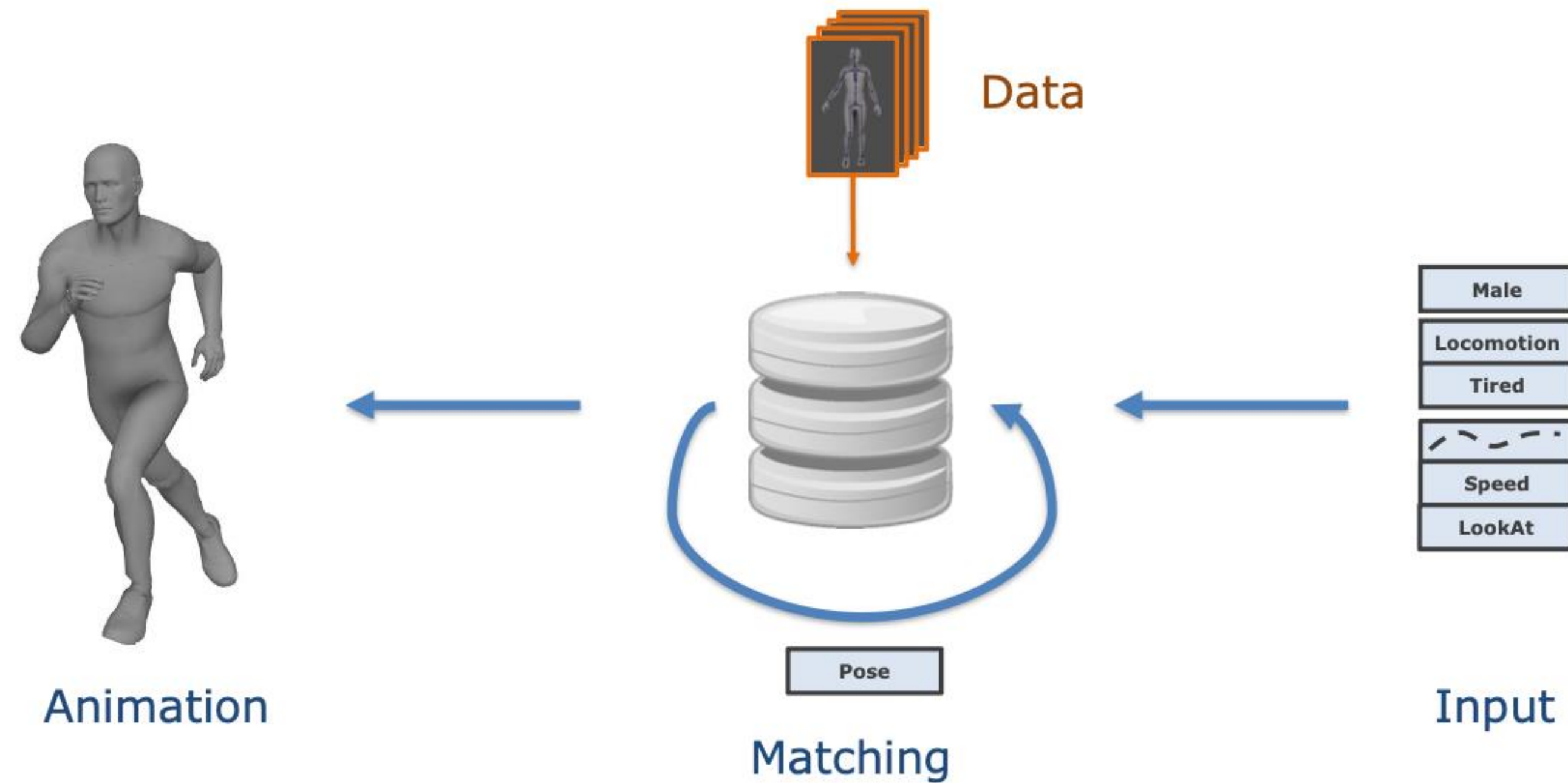
GDC talk by Daniel Holden

<https://www.youtube.com/watch?v=o-QLSjSSyVk>

PDF download: <https://www.gdcvault.com/play/1025389/Character-Control-with-Neural-Networks>



# Programming Details



Animation = Query(Dataset, variables)

Select **animation** where **variables=xxxxx** from **dataset**

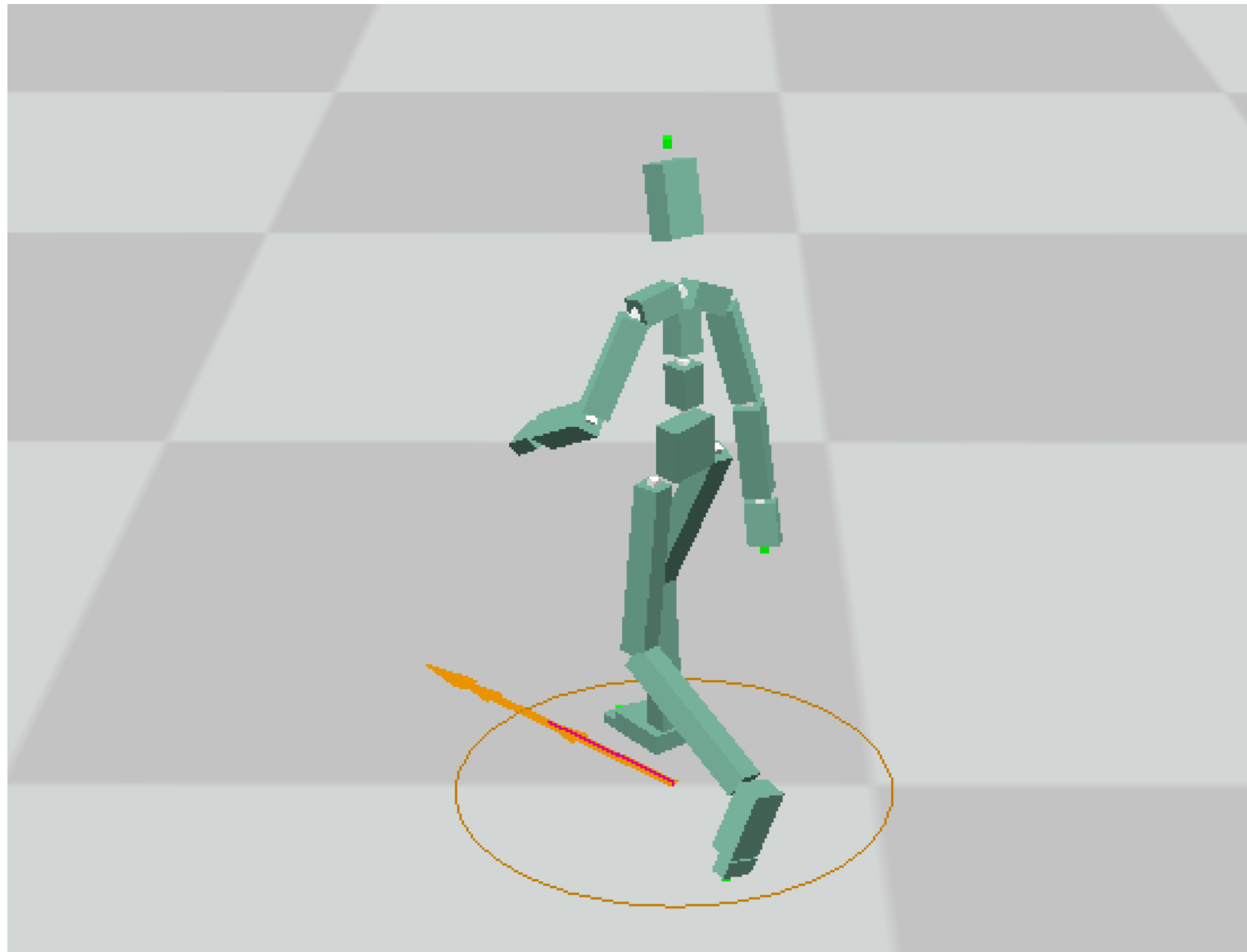
# Feature Design

For a motion sequence with  $n$  frames

Each frame has its own features:

- Position
- Rotation
- Positional/Angular Velocity
- Future Position/Rotation after 20/40/60 frames



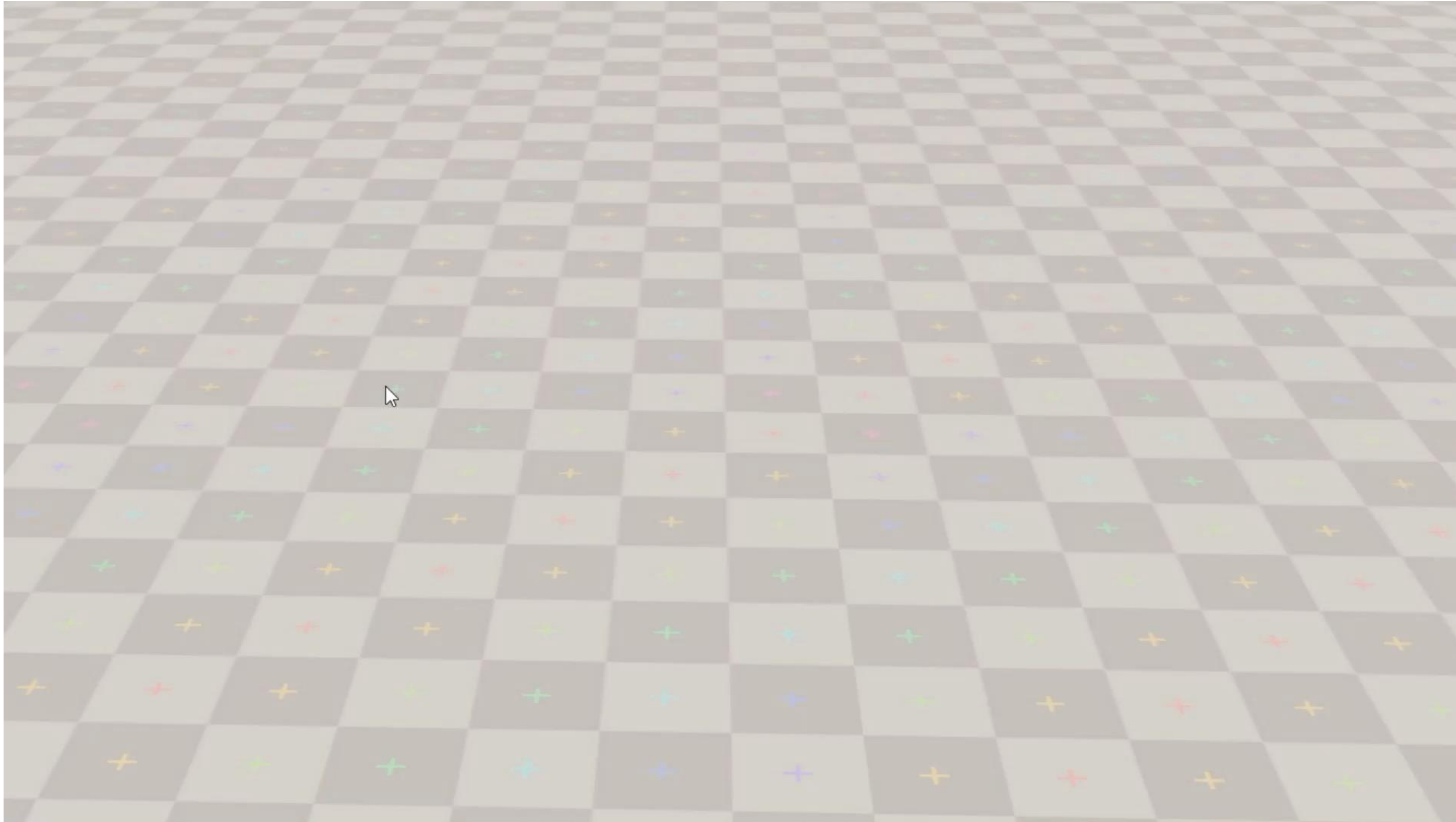


When we control the character

Each frame also has its own features:

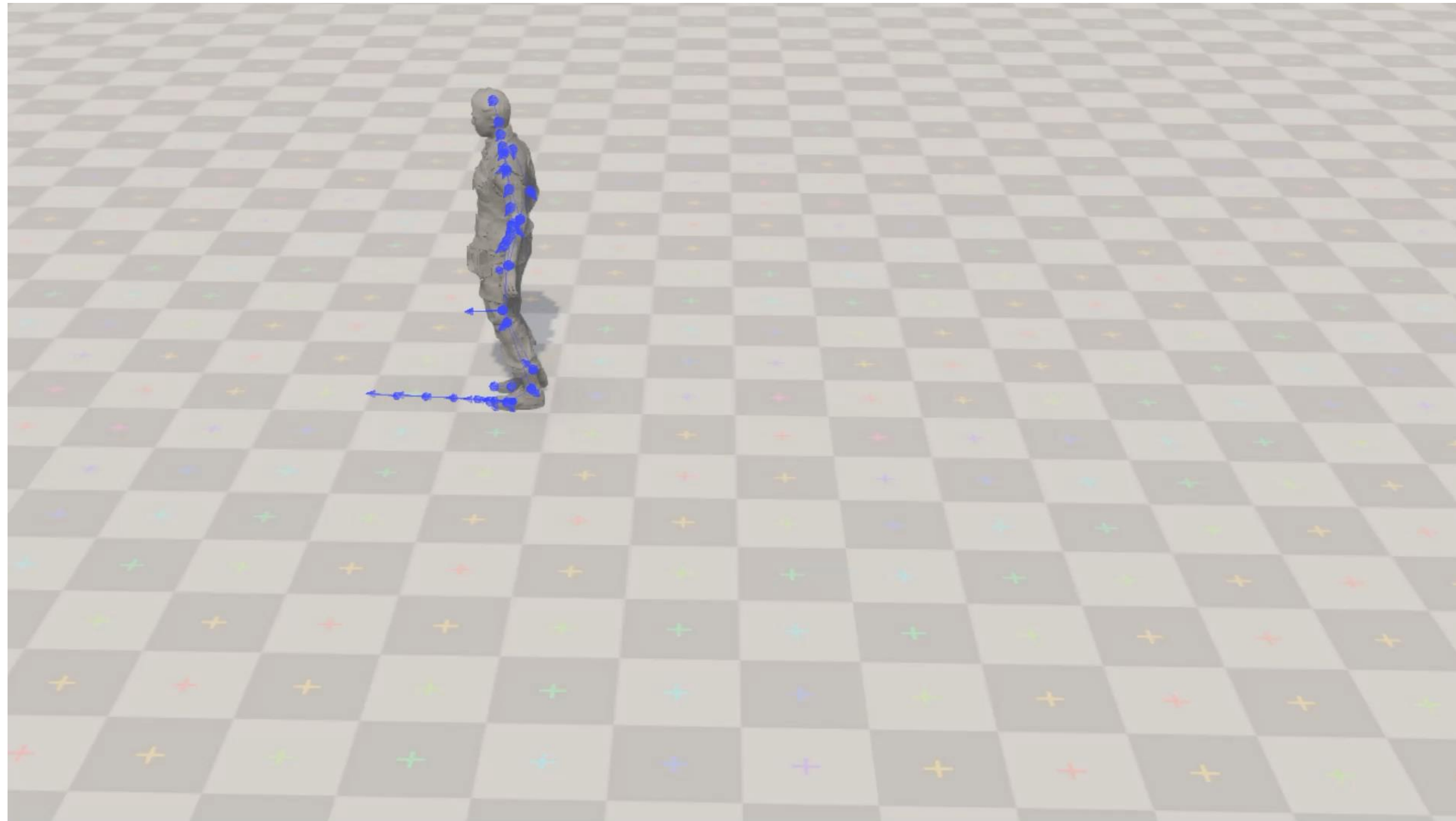
- Position
- Rotation
- Positional/Angular Velocity
- Future Position/Rotation after 20/40/60 frames (decided by controller)

# Step1: Give control signals





# Step2: Analyze the motion data



```
94 # Extract the data terms
95 pos, rot = forward_kinematics_with_channel(self.joint_parent, self.joint_channel, sel
96 rot = align_quat(rot, False)
97 vel = np.zeros_like(pos)
98 vel[1:] = (pos[1:] - pos[:-1])/self.dt
99 vel[0] = vel[-1]
100 avel = np.zeros_like(vel)
101 avel[1:] = quat_to_avel(rot, self.dt)
102 avel[0] = avel[-1]
```

We know:

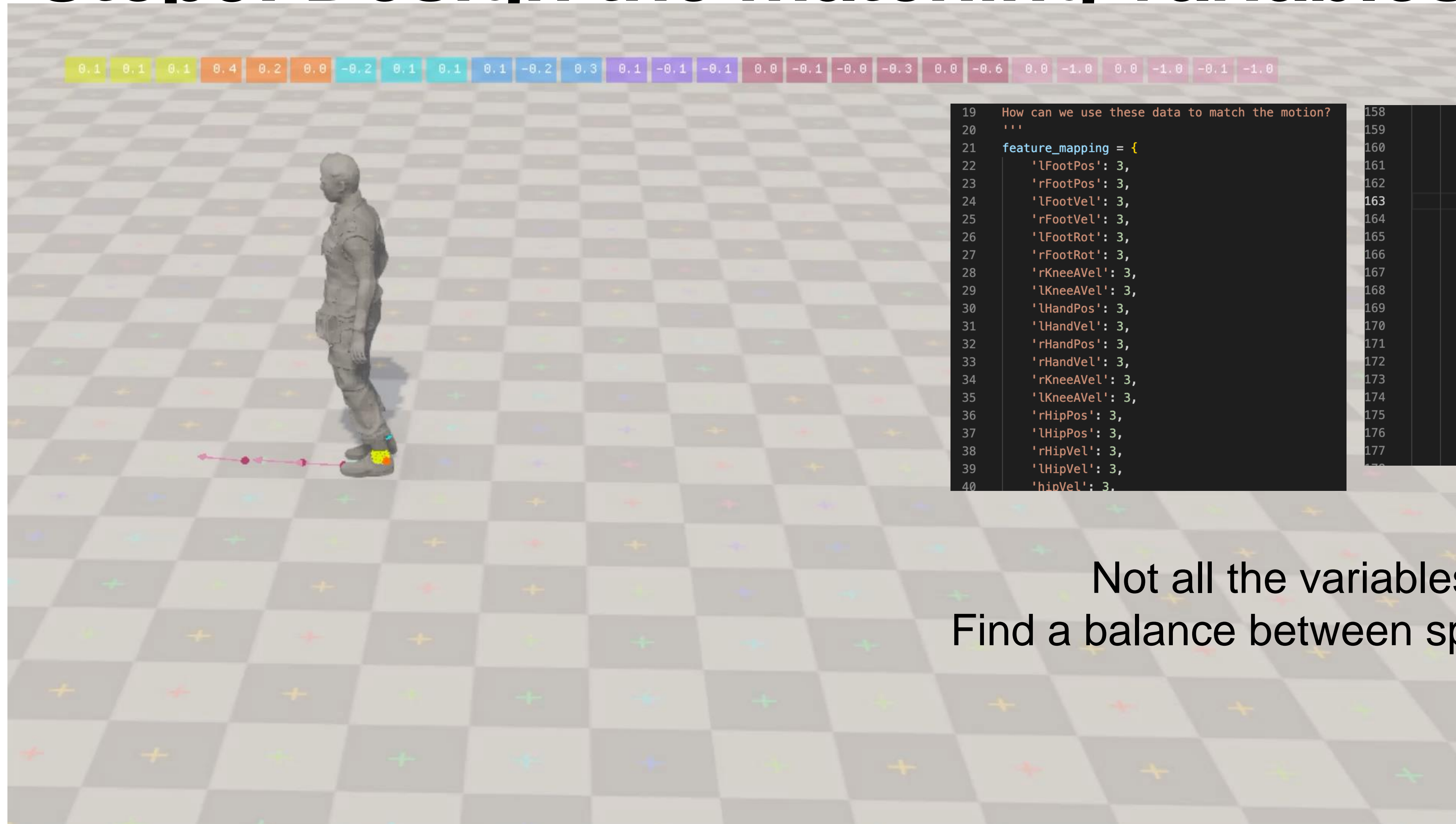
1. Joint position
2. Joint velocity
3. Joint rotations
4. Joint angular velocity

...

relative position  
waving timing

...

# Step3: Design the matching variables



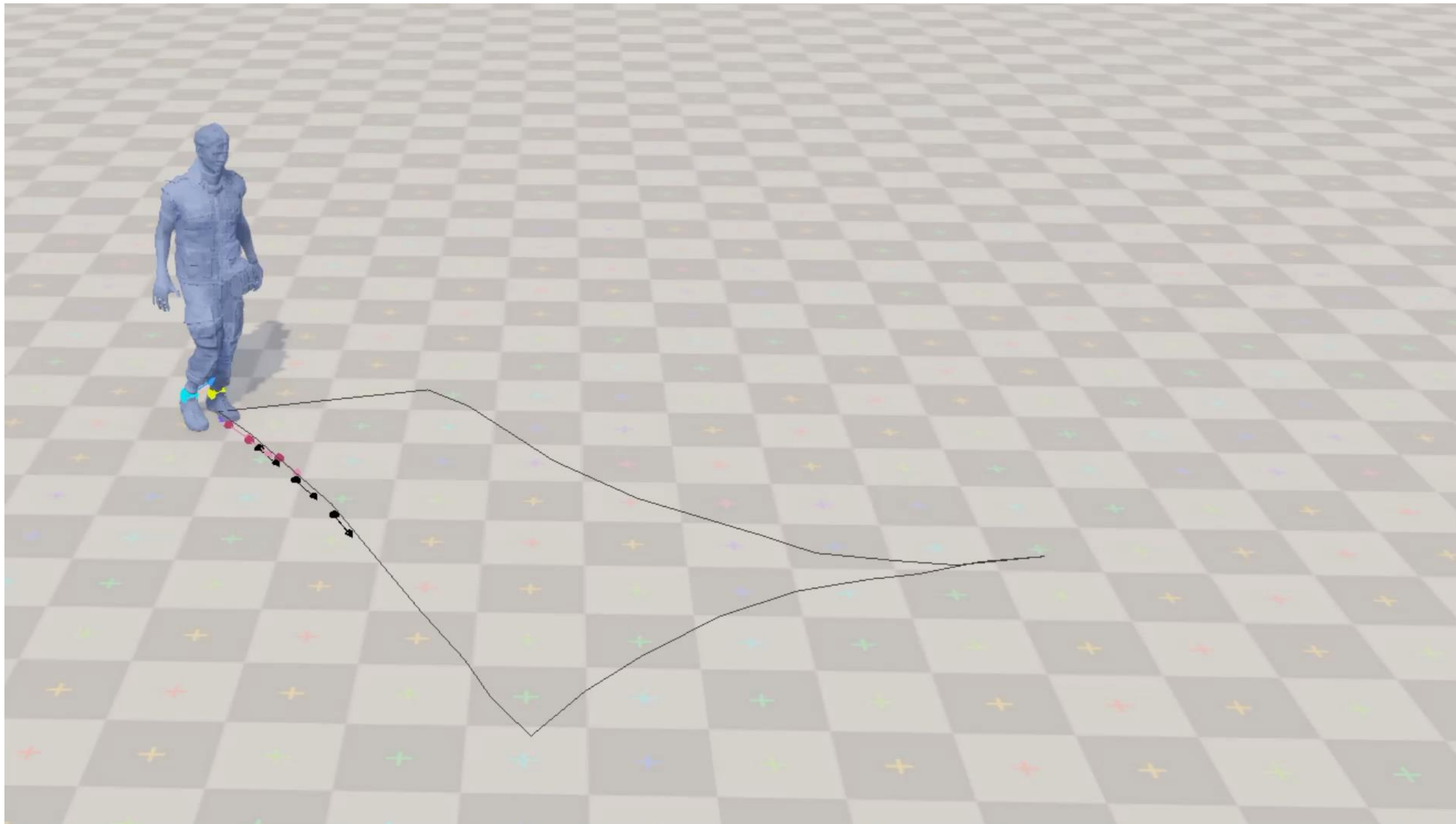
```
19 How can we use these data to match the motion?
20 '''
21 feature_mapping = {
22     'lFootPos': 3,
23     'rFootPos': 3,
24     'lFootVel': 3,
25     'rFootVel': 3,
26     'lFootRot': 3,
27     'rFootRot': 3,
28     'rKneeAVel': 3,
29     'lKneeAVel': 3,
30     'lHandPos': 3,
31     'lHandVel': 3,
32     'rHandPos': 3,
33     'rHandVel': 3,
34     'rKneeAVel': 3,
35     'lKneeAVel': 3,
36     'rHipPos': 3,
37     'lHipPos': 3,
38     'rHipVel': 3,
39     'lHipVel': 3,
40     'hipVel': 3,
```

```
158 for feature_name in self.feature_names:
159     '''
160     Extract the position:
161     features.append(self.extract_offset(root_r, feature_name))
162     Extract the rotation:
163     features.append(self.extract_rotation(root_r, feature_name))
164     Extract the velocity:
165     features.append(self.extract_vel(root_r, feature_name))
166     Extract the future position:
167     '''
168     ##### Code Start #####
169
170     if feature_name == 'lFootPos':
171         features.append()
172     elif feature_name == 'lFootVel':
173         features.append()
174     elif feature_name == 'trajectoryPos2D':
175         features.append()
176
177     ##### Code End #####
```

Not all the variables is needed  
Find a balance between speed and accuracy



# Step4: Matching



```
259     normalized_query = self.db.normalize_features(query_feature)
260
261     # Do the query
262     idx = self.db.query_tree.query(normalized_query.reshape(1,-1), k=1)[1][0]
```

Because we only use small dataset  
We also store it with OC-tree  
So the searching is fast

# Programming Details

Prepare:

- A motion dataset
  - Oc-tree to accelerate the searching
- A set of feature variables
- A animation system to play the animation

**The dataset and animation are prepared in this assignments  
you only need to care the variables design**



# What you need to do:

Design the matching feature and its weights

```
def main():
    viewer = SimpleViewer()
    controller = Controller(viewer)

    selected_feature_names = ['trajectoryPos2D', 'trajectoryRot2D']
    selected_feature_weights = [1, 1]

    # selected_feature_names = ['lFootPos', 'rFootPos']
    # selected_feature_weights = [1, 1]

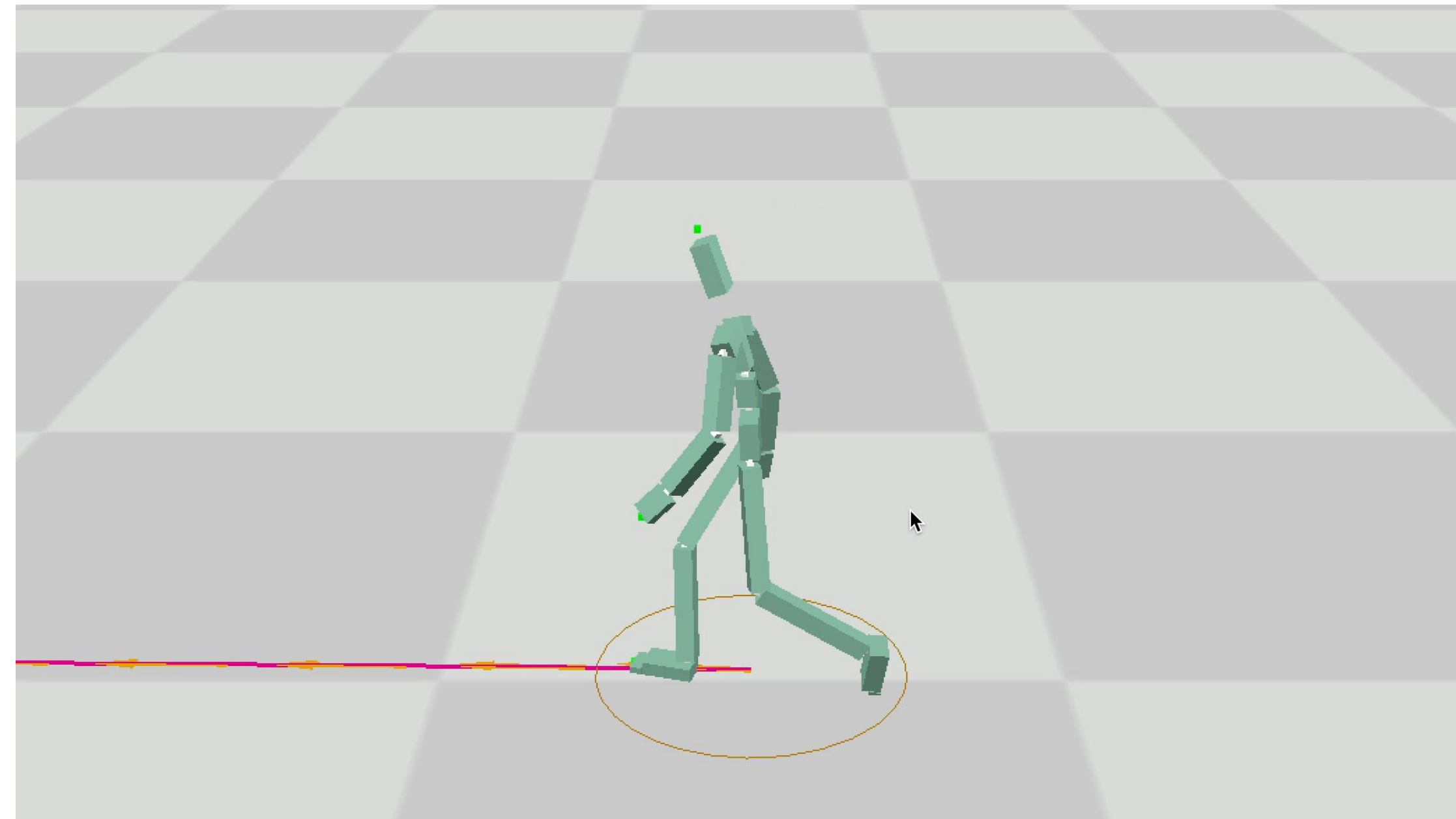
    assert len(selected_feature_names) == len(selected_feature_weights)

    character_controller = CharacterController(viewer, controller, selected_feature_names, selected_feature_weights)
    task = InteractiveUpdate(viewer, controller, character_controller)
    viewer.addTask(task.update)
    viewer.run()
    pass
```

- Some features are pre-defined. (L11~L35, L135~L197)
- If you want to add you new features, update the code around (L11~L35, L135~L197)

# Desired Results

1. Less variables, and better performance (total 15%, 22% - your\_variable\_num)
2. System analyzation (10%) about variable selection, future frame range, etc.



# Overview, Due: Nov. 12

- part1\_key\_framing (30%)
  - - Linear interpolation (10%); Slerp Interpolation (15%)
  - - Report the different performance by giving different numbers (5%)
- part2\_concatenate (35%)
  - - Define the search window (10%); Calculate the sim\_matrix (10%); Find the real\_i and real\_j (10%); The shifting on the root joint position (5)
- part3\_motion\_matching (25%)
  - variable terms (15% <- 22% - your\_variable\_num) + System analyzation (10%)
- Report (8%) + 2 videos (2%)
  - Including necessary experiment results by **different parameters (4%)** and **your thinking(4%)** for how to produce high quality motions