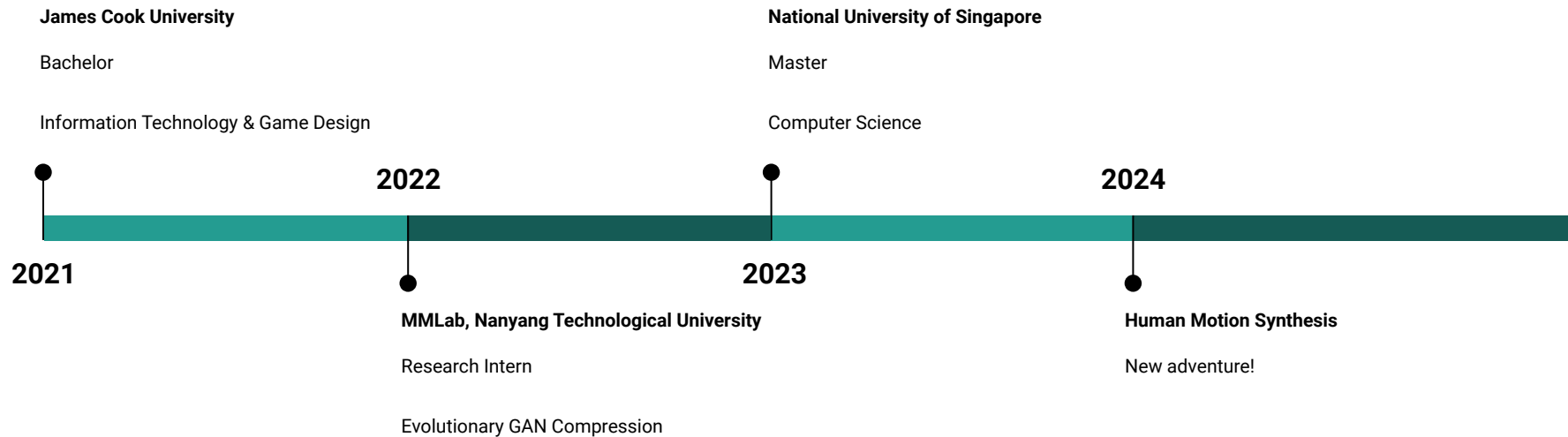


Atomic Motions

Towards bottom-up motion synthesis for digital humans

Self Introduction



What creates motion?

Motivation

Shaolin Temple, 18 km from my city

I always wanted to learn kung fu...
until I discovered computers.

How to do both?

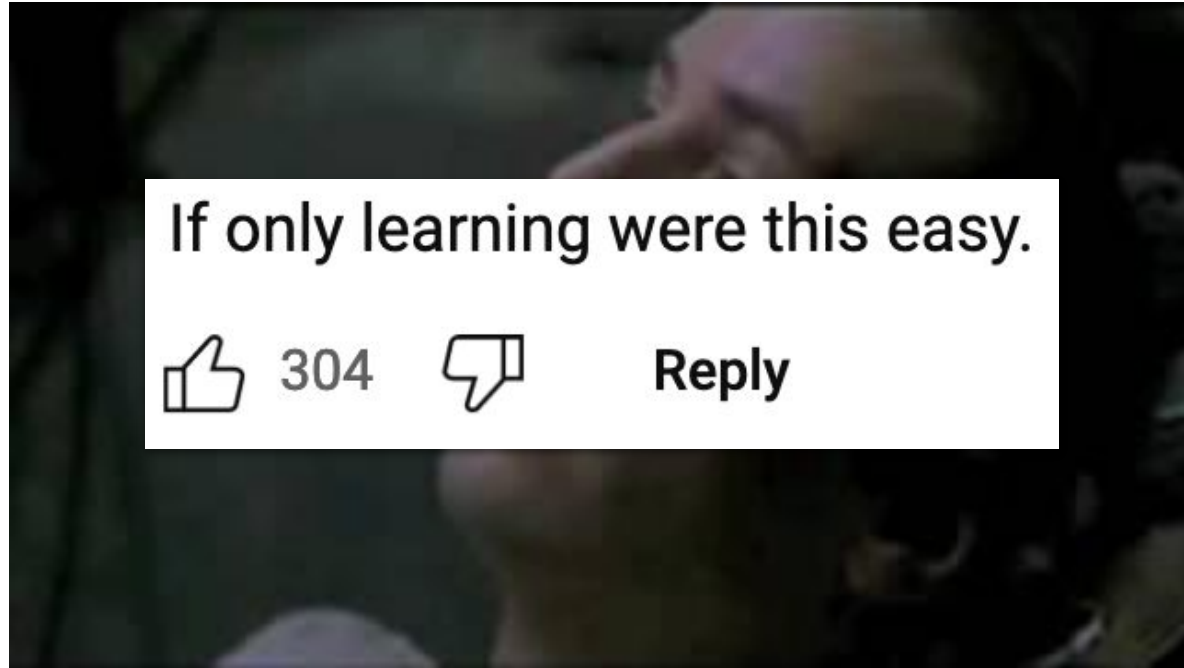


Motivation



“I know kung fu.” (His digital human does.)

Motivation



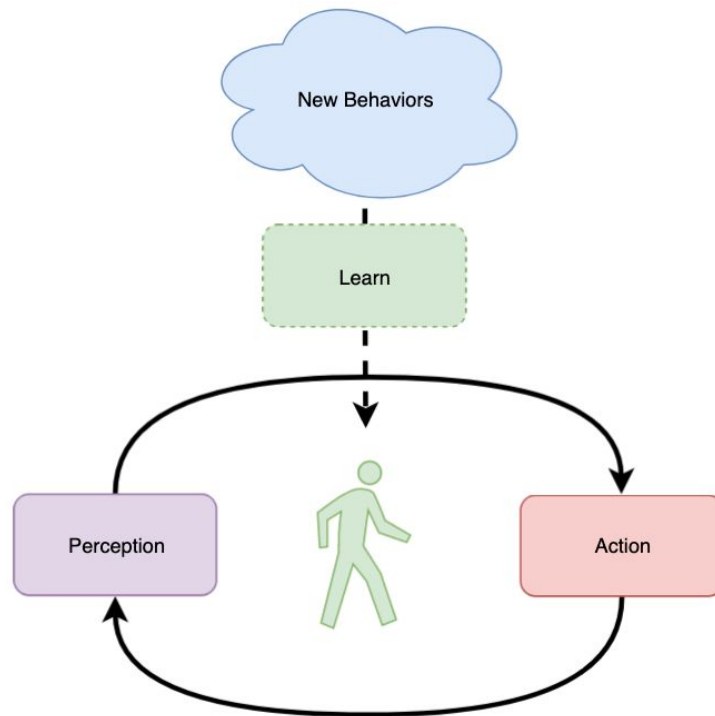
Top comment of “I know kung fu.”

Motivation

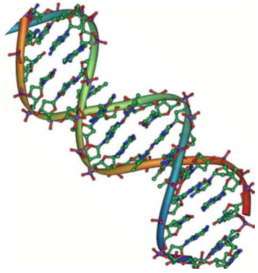
Digital humans should perform
human-like behaviors like kung fu.

If not, can only call them “puppets”.

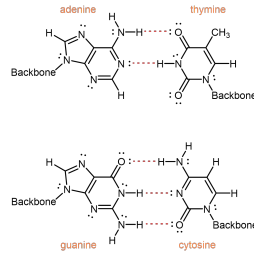
The basic behavior is motion.



What creates motion?



DNA



Nucleobases

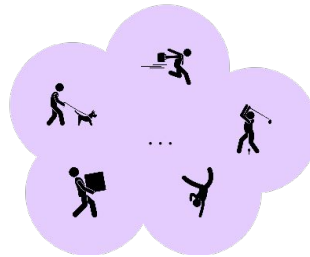
The quick brown fox
jumps over
the lazy dog.

Sentence



quick brown
fox jump
lazy dog

Words

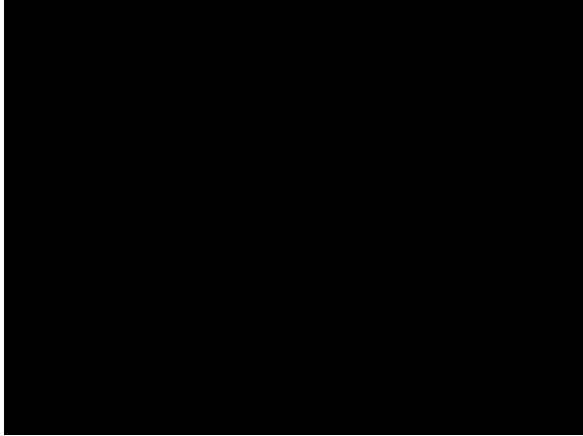


Motion



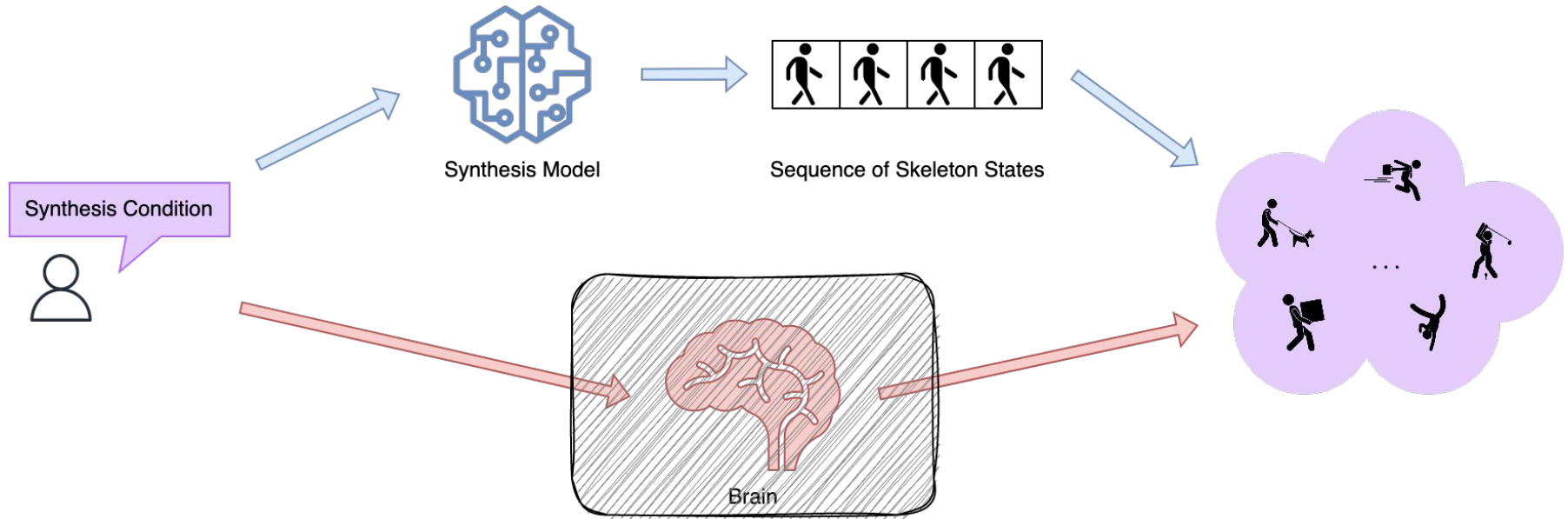
What creates motion?

We know how to capture and represent motion.



What creates motion?

Produce skeleton states instead.

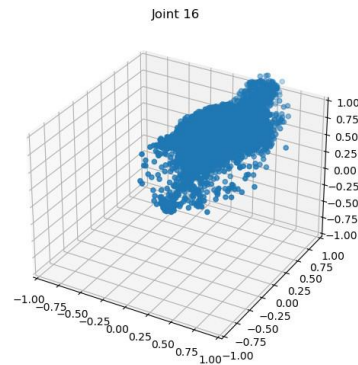
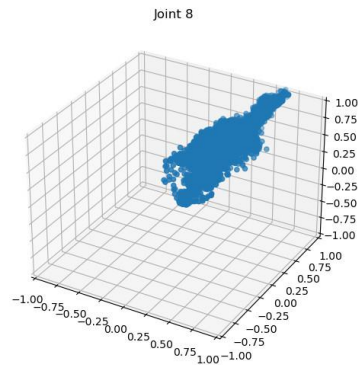
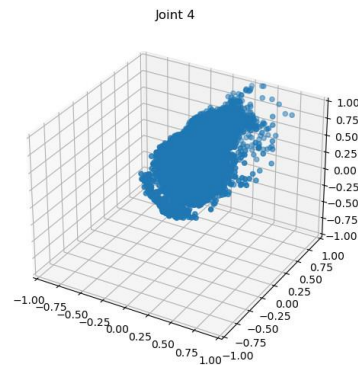
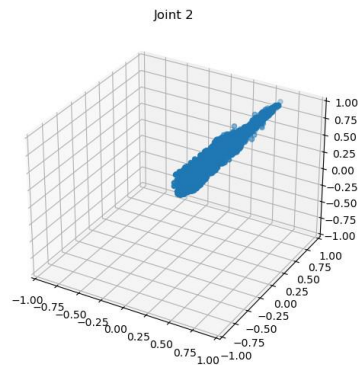


What creates motion?

Human motion occupies a very small portion of skeleton state space.

Weeding out this space is challenging.

What about humans?



What creates motion?

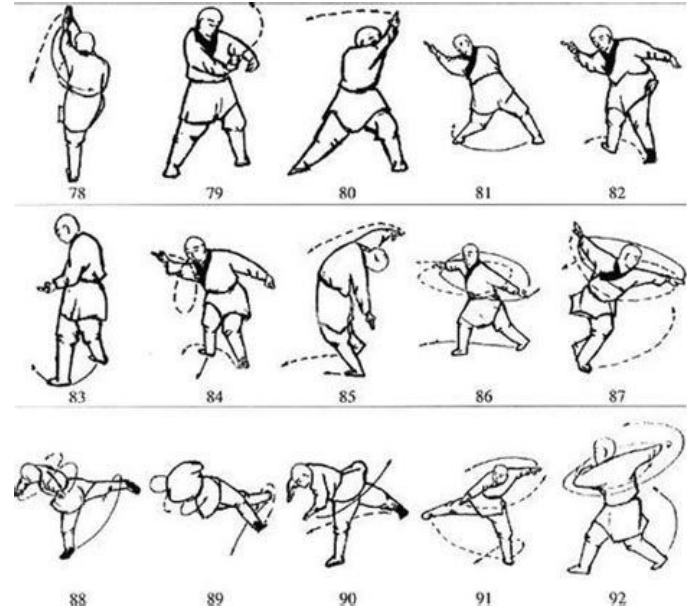
Humans learn how to move from

very small data

108 illustrations

=> 3 hours of performance

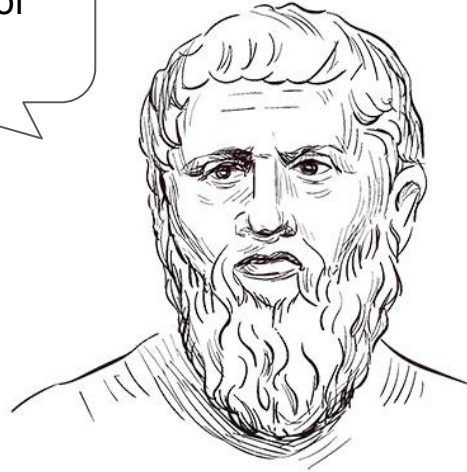
=> 129600 frames of 12Hz animation



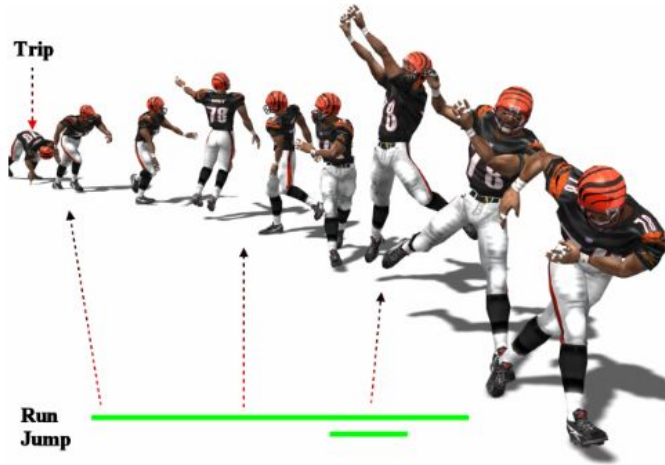
What creates motion efficiently?

Plato

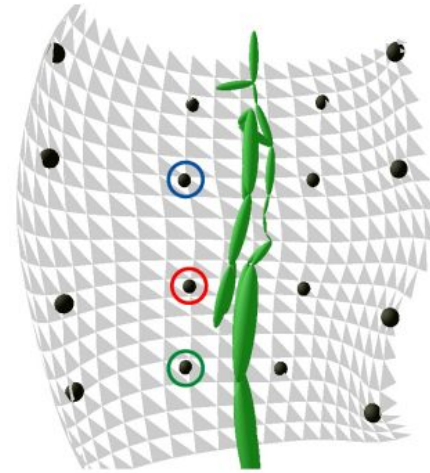
Self-motion is the very
idea and essence of
the **soul**.



2000s: Motions as Editing Templates

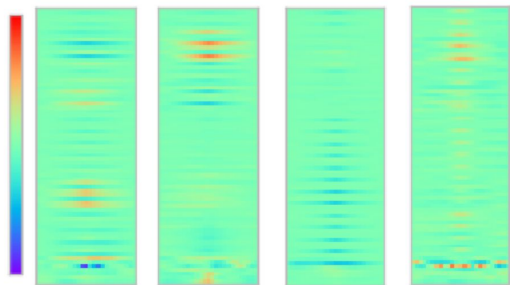


Dynamic programming to **select** from template motion snippets with a cost function.



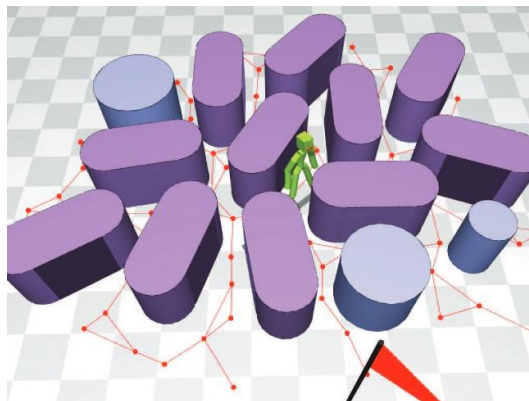
Interpolate template motion snippets (black dots) with kriging to reach new states (other vertices).

2010s: Motion Manifolds and other Goodies



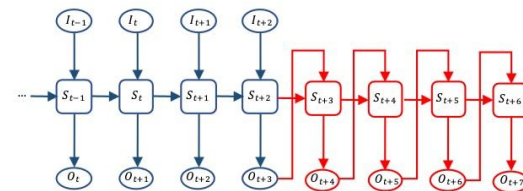
A **motion manifold** constraining a motion's degrees of freedom along time.

“Motions as low-dim features.”*



Generating squeeze-through motions via **path exploration**.

“Motions as path optimization.”

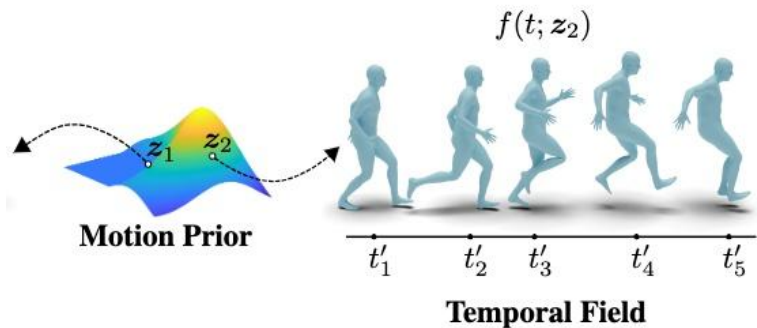


Predicting **motion transitions** in a Markov chain.

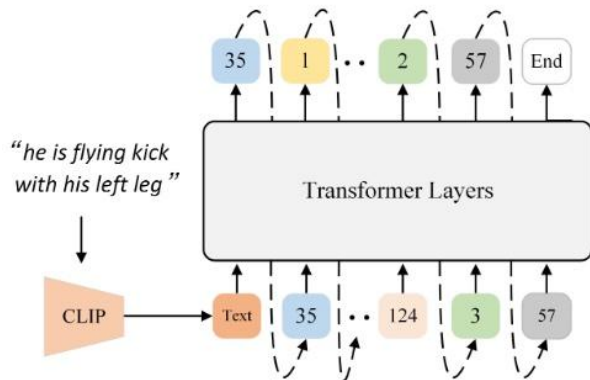
“Motions as state transitions.”

*Linearly interpolatable low-dimensional states residing on a high-dimensional manifold.

2020s: Generative Manifolds and NLP

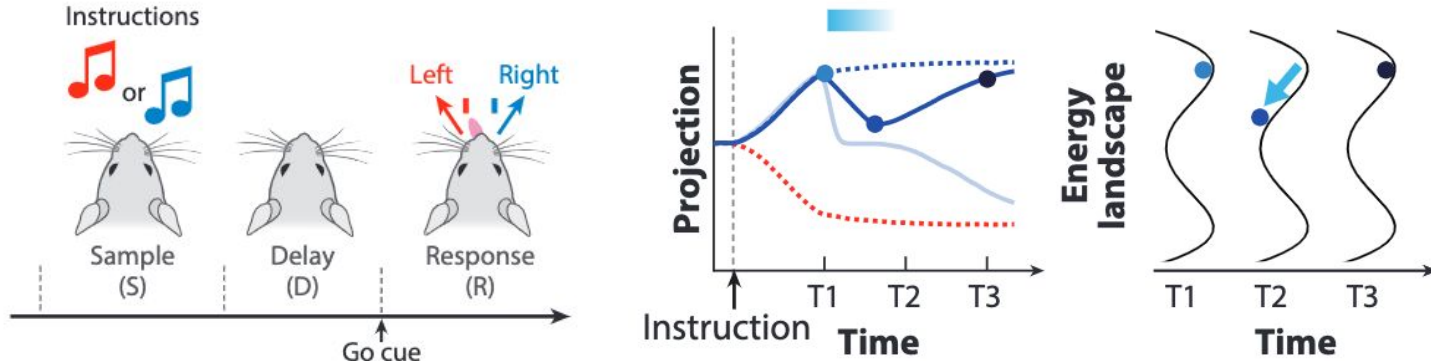


Motion manifolds become probabilistically described, or allowing noise injection.



Others treat motion as a language and **translate** synthesis conditions to it.

2020s: Neural Dynamics and Attractor Hypothesis



Exciting work in neuroscience: perturbation experiments change mice motions and supports attractor hypothesis:

Motion states are **planned** through a dynamical system.

Attractors are stable states encoding well-trained motions.

Top-down or Bottom-up?

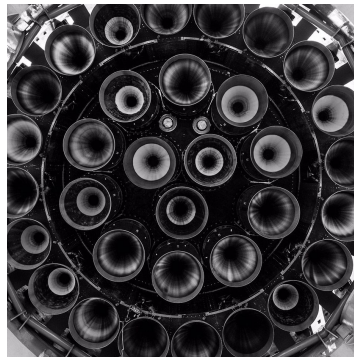
From template editing to motion manifolds, **lower-level** building blocks are being explored.

Many advantages:

- Data and Computational Efficiency
- High-frequency Motion Details
- Motion Diversity



RS-25 on Saturn V



Raptors on
SpaceX Starship

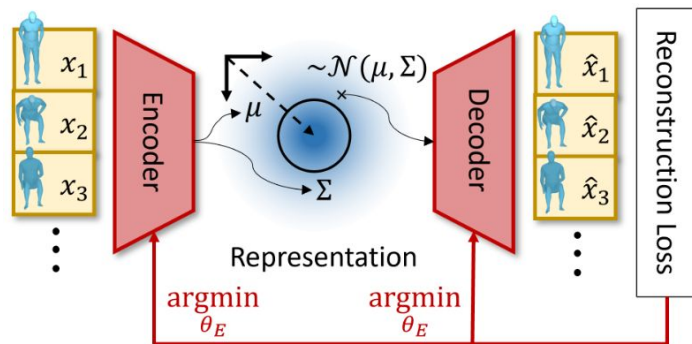
“There is plenty of room at the bottom.”
- Richard Feynman

Atomic Motions

The Existing Bottom-up Hypothesis

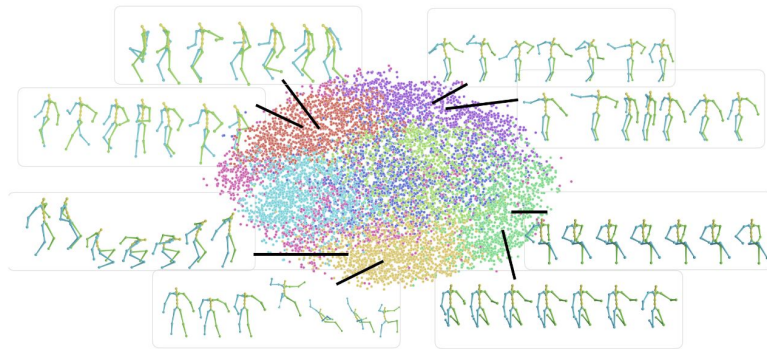
Building blocks construct a **uniform, single-level** structure for all motions.

- Often a hypersphere or multivariate normal.



Low distinction between building blocks.

- A uniform set of **affinities**.
- A uniform set of **aversions**.



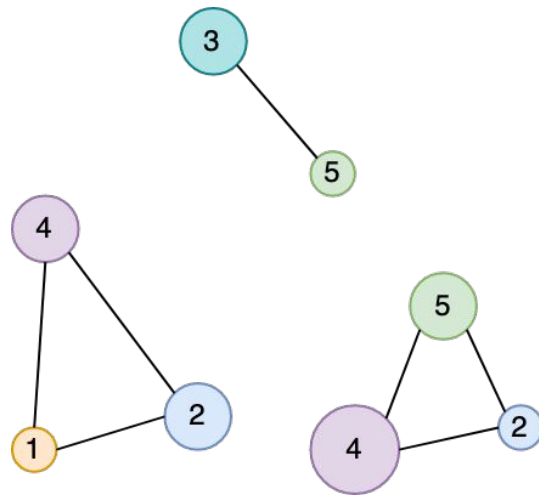
My Hypothesis

Atomic motions construct **distinct, hierarchical** structures to form **distinct** motions.

Each pair of atomic motions

- exhibits unique affinity.
- exhibits unique aversion.

A small set of atomic motions suffices to provide diverse motion structures.



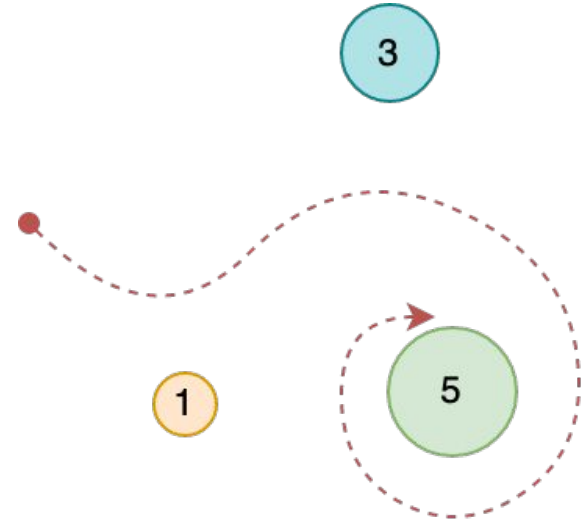
My Hypothesis

One additional assumption for **motion planning**:

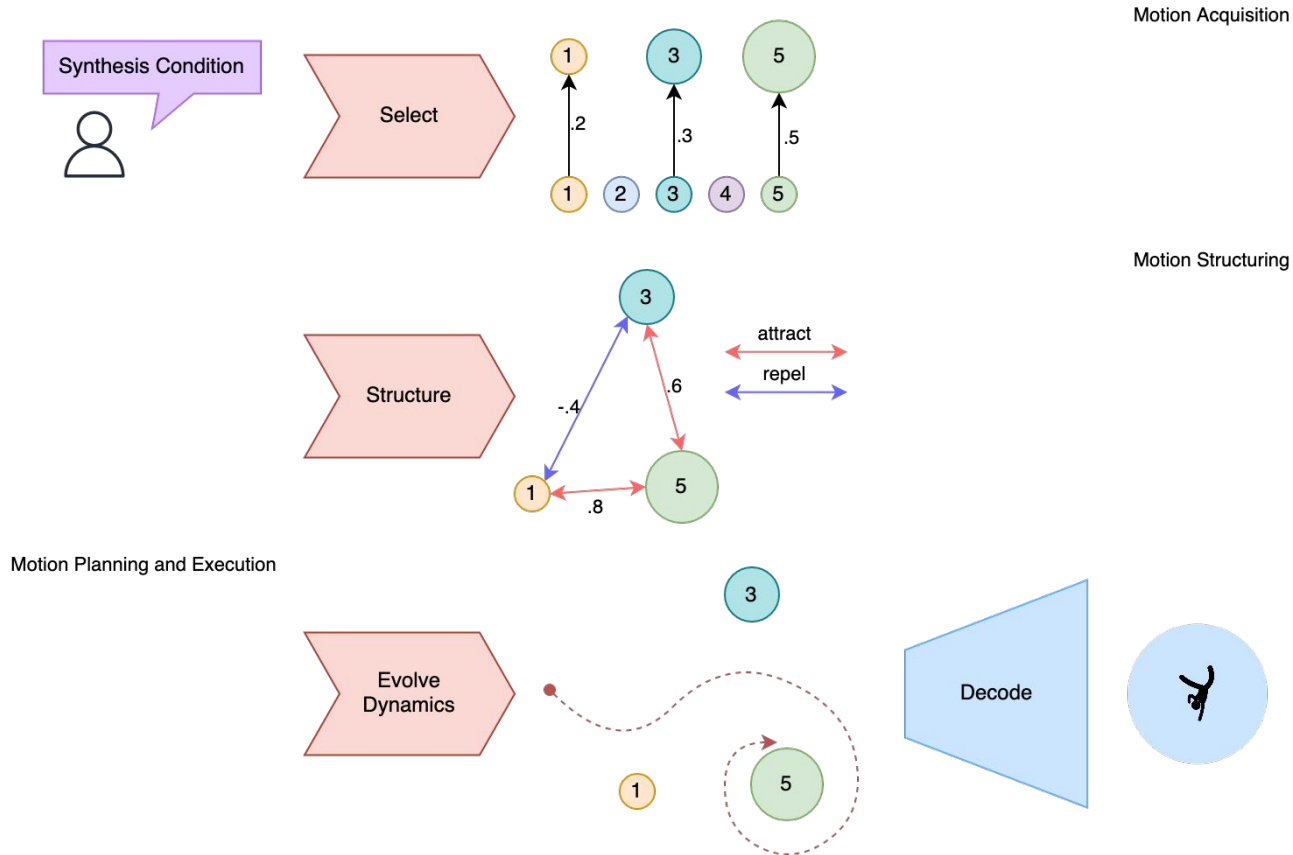
Given planning duration and motion structure,
a motion state is evolved in a
nonlinear dynamical system.

The structured atomic motions are **stable states**.

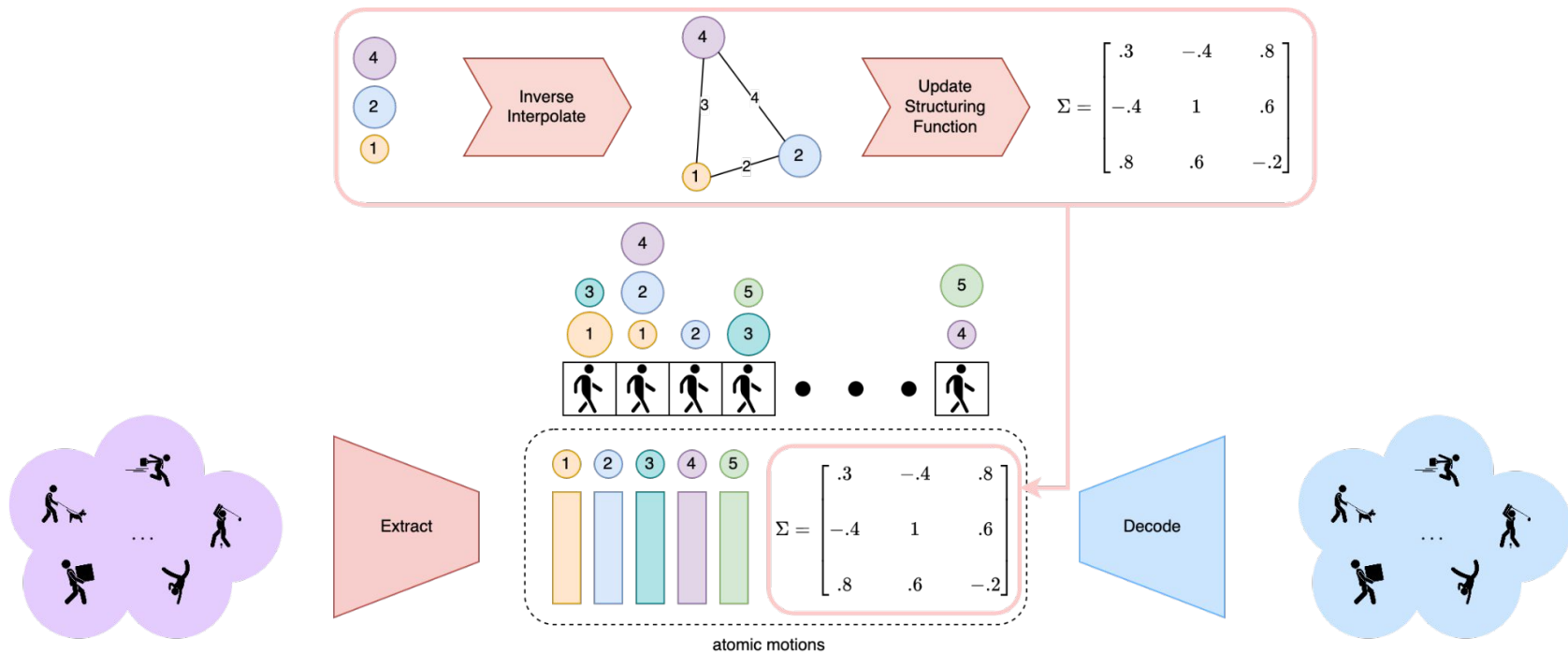
This assumption allows for **optimal motion control**.



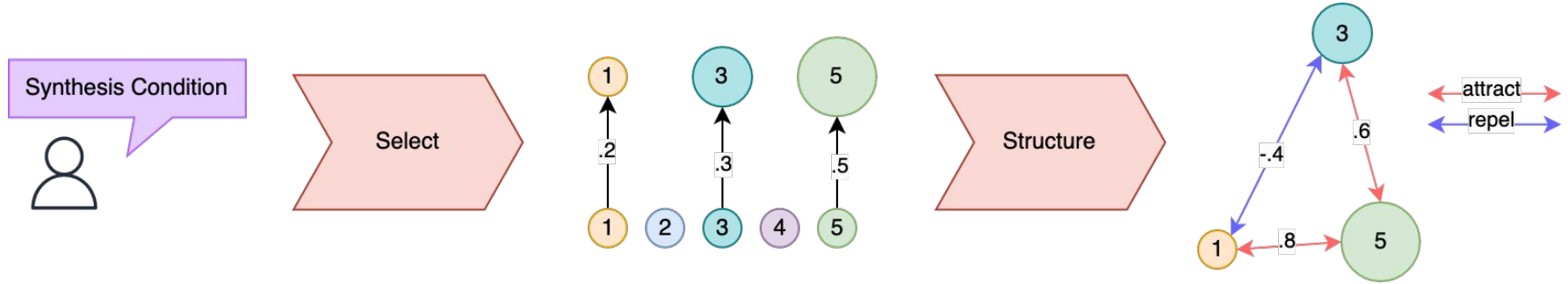
Overview of the Atomic Motions Framework



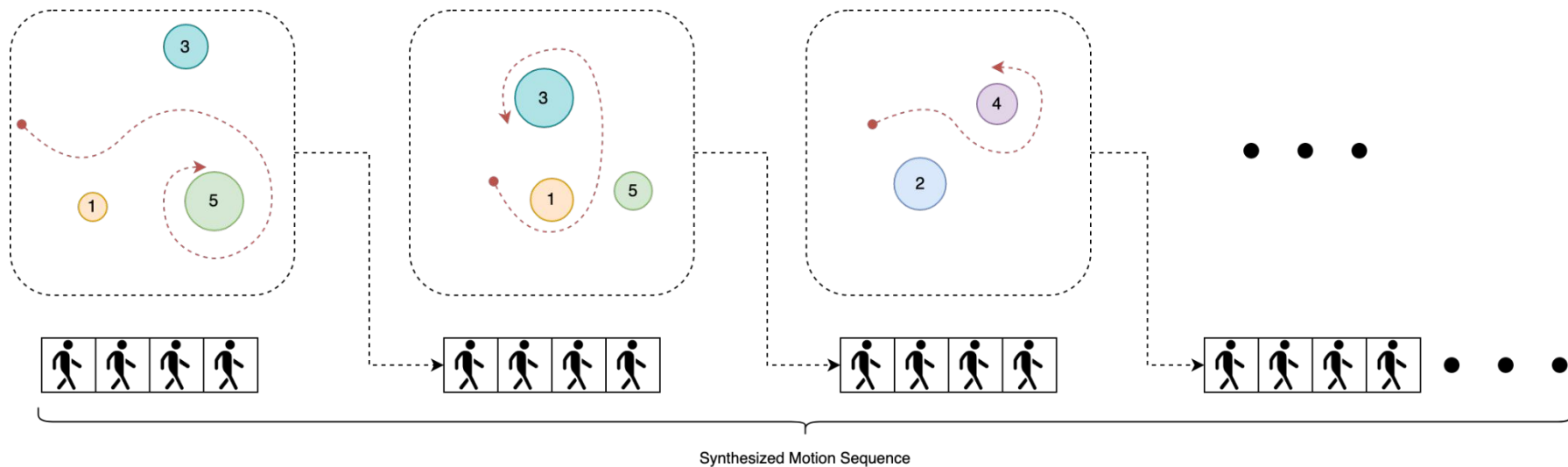
Pipeline 1: Extraction Stage



Pipeline 2: Selecting and Structuring Stages



Pipeline 3: Planning and Execution Stages



Grow structures and control dynamics
with atomic motions.

Growing Structures

Amino acids grow into a protein structure.

Atomic motions grow to form complex motion structures.

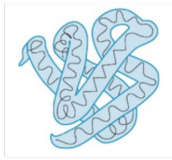
Top structures are always built with **plausible** low structures.

Synthesis conditions can be propagated down **without looping**.

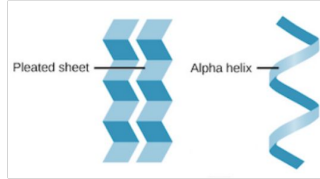
Quaternary Structure: 3D Structure from Multiple Sequences



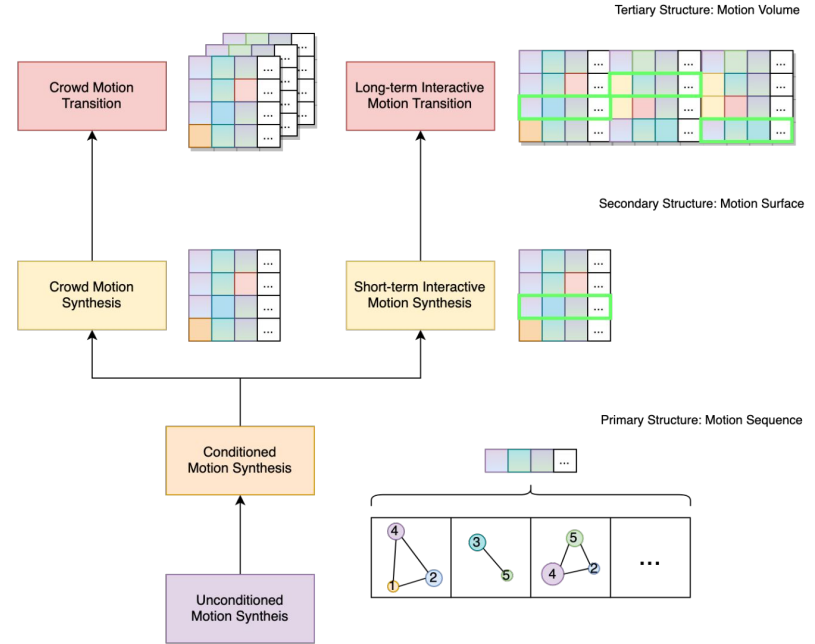
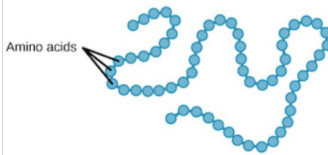
Tertiary Structure: 3D Folding Patterns



Secondary Structure: Repeating 2D Patterns



Primary Structure: Amino Acid Sequence



Controlling Dynamics

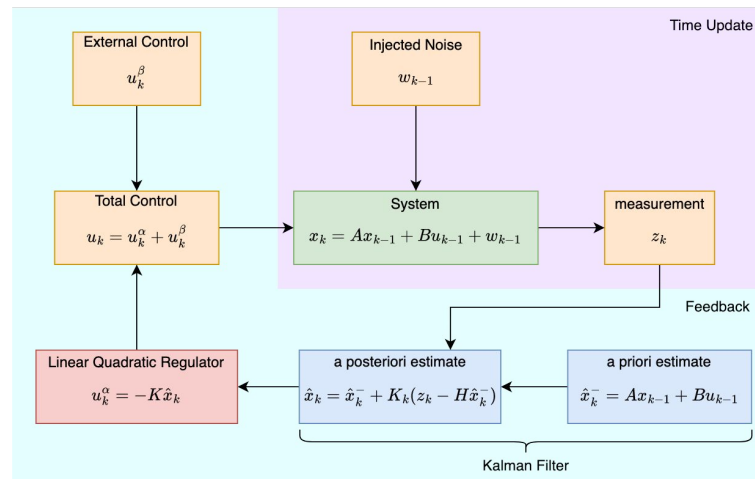


We want diverse plan dynamics between two motion states.

Disturbance noise is injected only in the middle.

Data-driven methods can't denoise the motion signal well to plan the next move.

We can get rid of it in our framework with Kalman filtering and optimal feedback control.



Controlling Dynamics: Motion Processing

Apply different **filters** in spectral domain of a plan trajectory to model different **motion styles**:

- FFT Components => Style Transfer
- Convolution => Smooth and Sharpen
- Optimal Filtering => Denoising and Compression

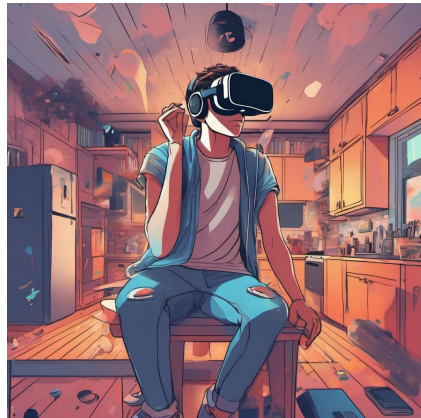
Nurturing data-efficient digital humans.

Data-efficient Digital Humans

Motion acquisition becomes **cost-effective**: atomic motions can be extracted from **small data**.

Computational cost also drops, meeting realtime needs in **VR** and **crowd simulation**.

Online learning is possible: atomic motions can be acquired through active sampling for novel tasks.



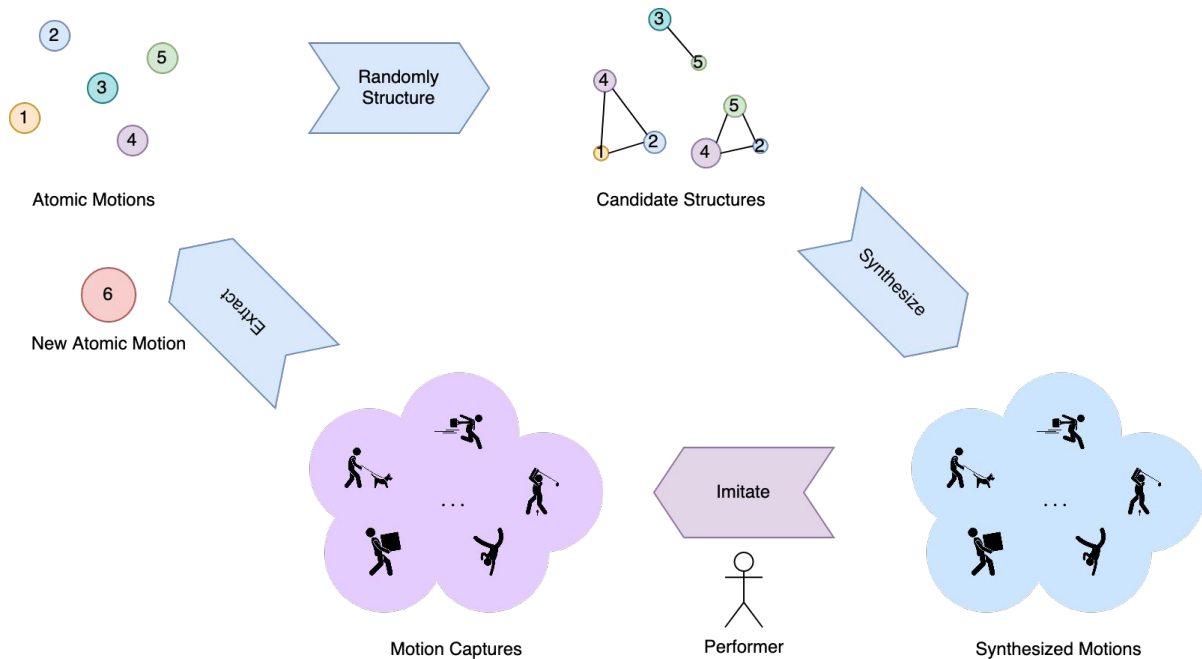
Meet challenges and limitations.

Data Collection

Essential motion features are hard to pick up if they are **outliers** in data.

Active sampling can help, but incurs human labor.

Its pipeline is also nontrivial to design and implement.



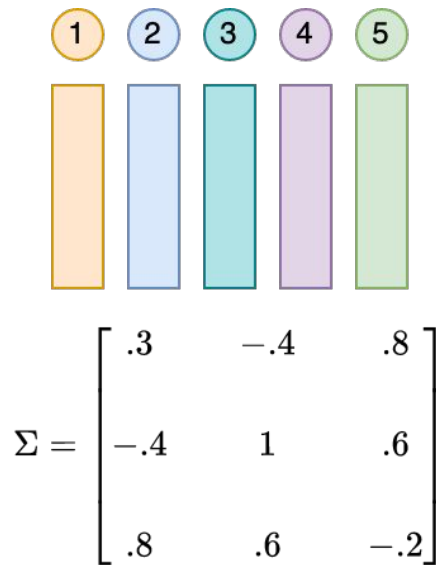
Information Quantification

How much information should we extract from a dataset?

- How many atomic motions?
- How long is each feature?
- How sparse is Σ ?

Currently, can only answer with tedious **ablation studies**.

Better option is **meta learning** to optimize quantification parameters across synthesis tasks.



Motion Integration

Atomic motions only produce **body motions**. Eventually, we need to integrate with other partial motions:

- Face motions.
- Hand motions.

Distinct **correlations** between them must be respected.

Atomic motions are too low-level to match with. Its higher-level structures are combinatorial.

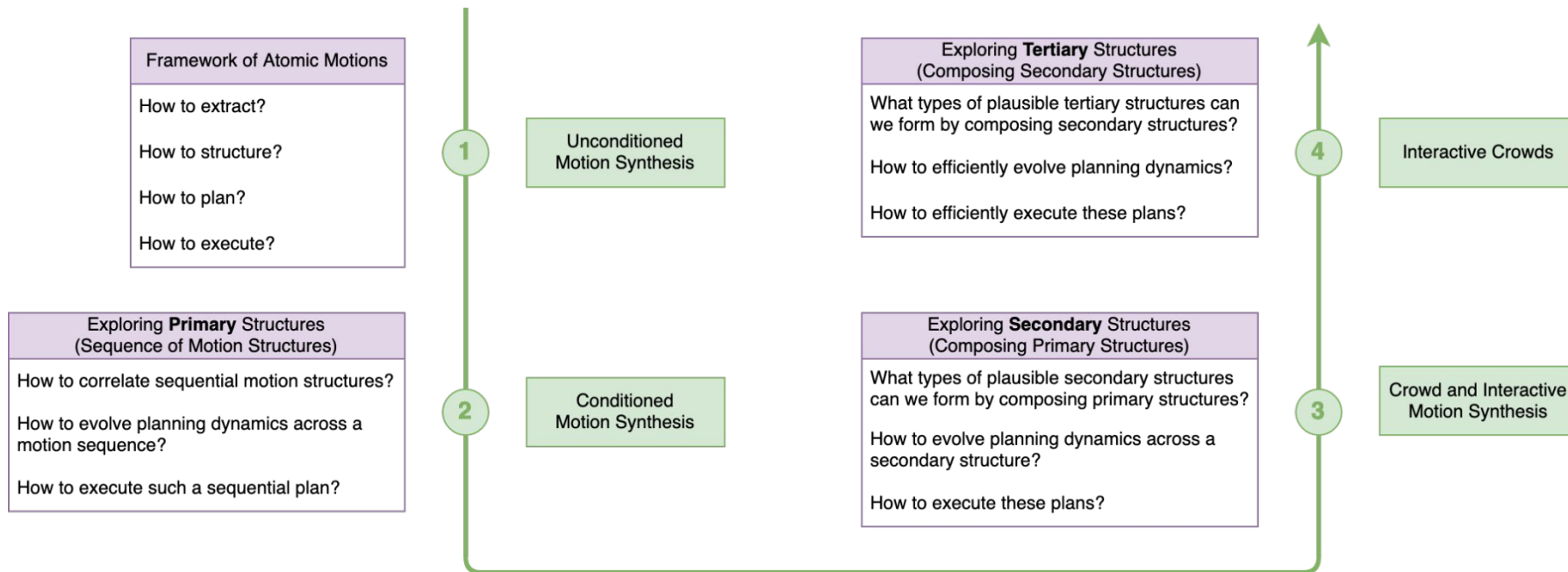
It's ill-posed to observe these correlations from data.



Research Forecast

I want to nurture self-driven digital humans.

Research Forecast



Thank you. Questions?

Contact

Wei Guo

Email: wei.parker.guo.pb@gmail.com

Site: <https://wei-parker-guo.github.io/>

A much detailed version of this presentation is at
https://wei-parker-guo.github.io/atomic_motions

(Alternatively, scan the QR code.)

