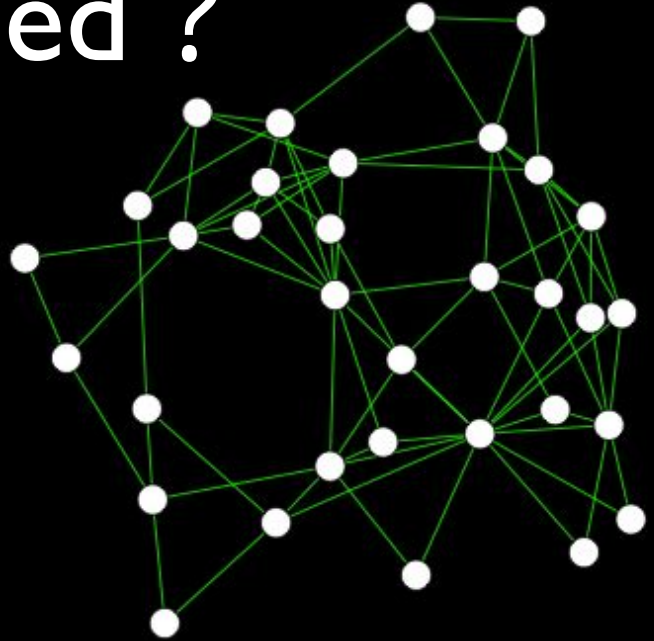
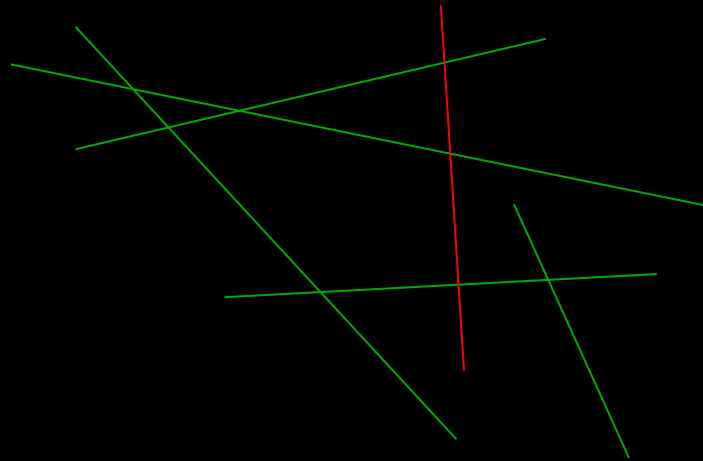


How can human movement be modelled ?

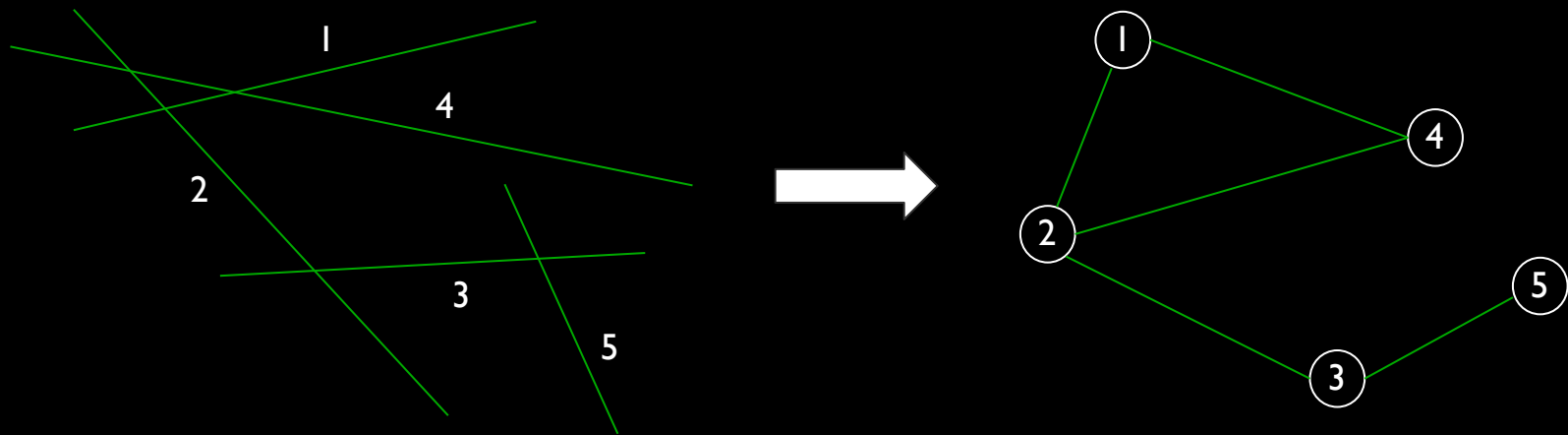
Soma Suzuki



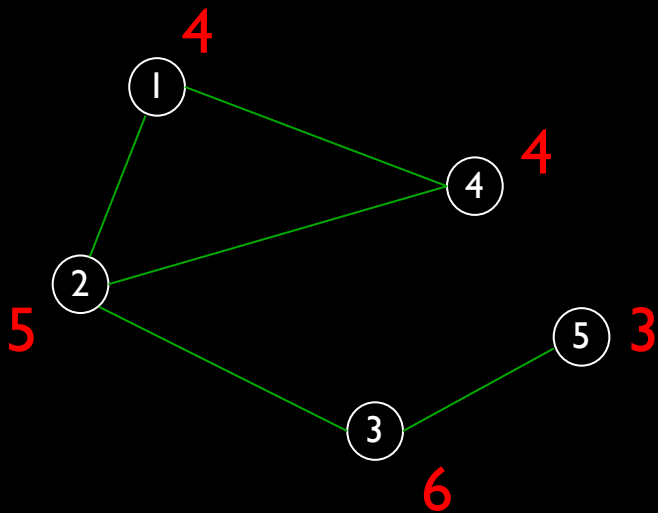
Context



Context



Context

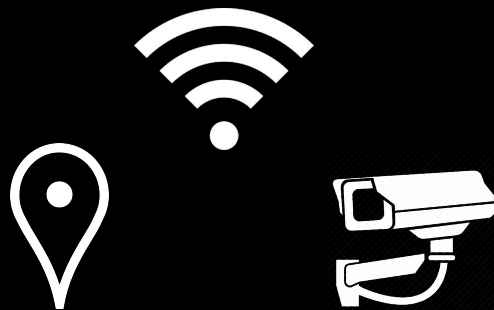
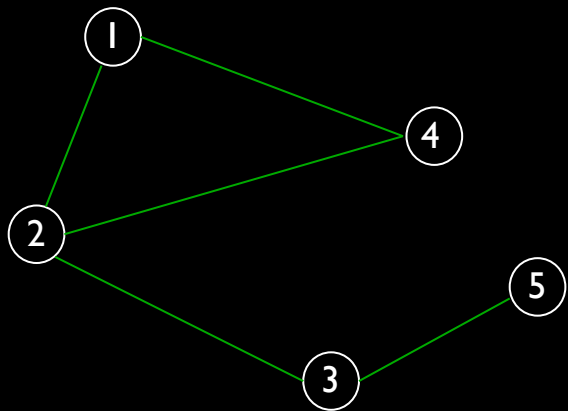


(Jiang '08)

$$\sum_{s=1}^k s \times N_s$$

$$\textcircled{2} \ 3 \times 1 + 1 \times 2 = 5 \text{ (if } s = 2\text{)}$$

Let's learn from human



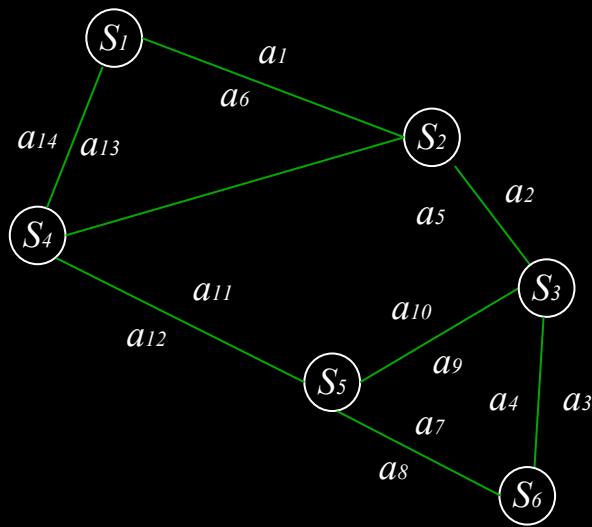
Imitation Learning



Expert trajectory dataset

$$D = \{\zeta_1, \zeta_2, \dots, \zeta_m\}$$

Markov Decision Process



MDP : tuple (S, A, P, R, Γ)

- $s \in S$: set of states
- $a \in A$: set of actions
- P : state transition matrix
- R_{θ} : reward function
- $\Gamma \in [0, 1]$: discount factor

Maximum Entropy Inverse Reinforcement Learning

(Ziebart '08)

$$D = \{\zeta_1, \zeta_2, \dots, \zeta_m\}$$

$$P_{\theta}(\zeta) = \frac{1}{Z} \exp(R_{\theta}(\zeta))$$

$$\max_{\theta} \underbrace{\sum_{\zeta \in D} \log P_{\theta}(\zeta)}_{L(\theta)}$$

Maximum Entropy Inverse Reinforcement Learning

(Ziebart '08)

Algorithm

$\theta_0 \leftarrow \text{RewardParameterInitialization}$
for $i = 1 : N$

$R_i \leftarrow \text{UpdateReward}(\theta_i)$

$\pi_i \leftarrow \text{UpdatePolicy}(R_i)$

$P \leftarrow \text{UpdateProbabilityDistribution}(\pi_i)$

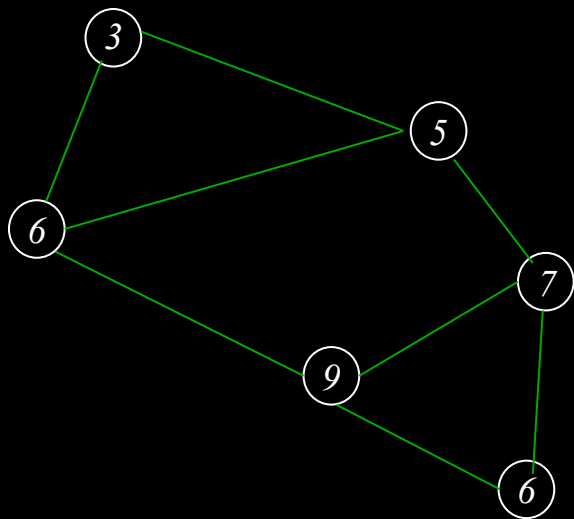
Compute $\nabla L(\theta)$

$\theta_0 \leftarrow \text{UpdateRewardParameter}(\nabla L(\theta))$

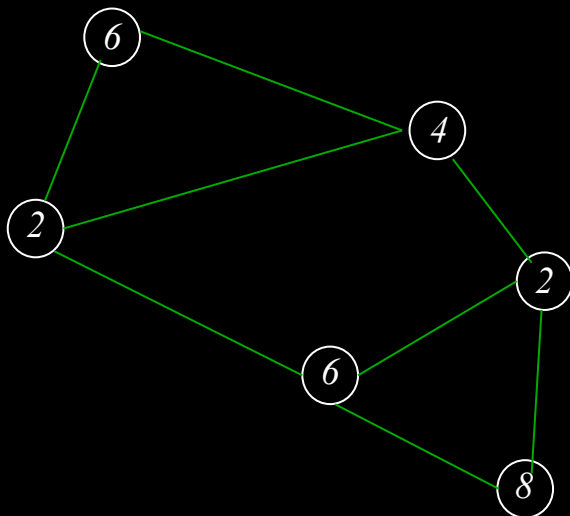
Evaluation

↔
Correlation

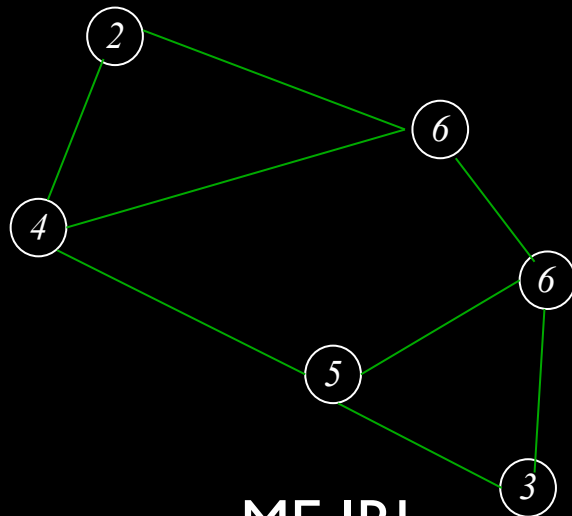
↔
Correlation



Space Syntax



Human Movement



ME IRL

Problem that I need to overcome



Collecting data pertaining to individual walking paths is very challenging (Torrens '10)

Questions?