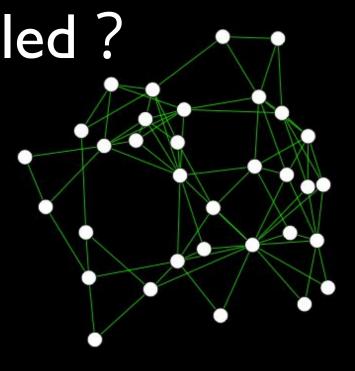
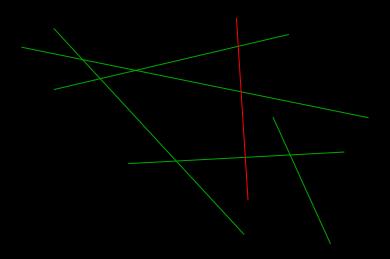
How can human movement be modelled?

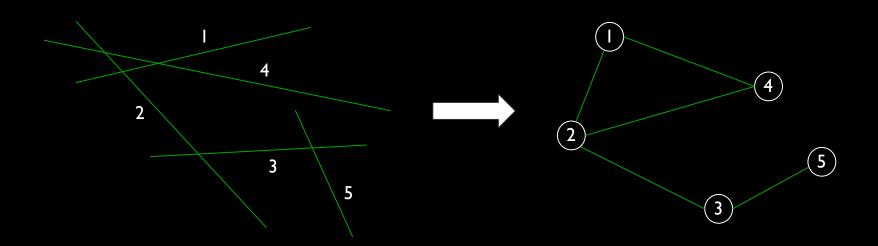
Soma Suzuki



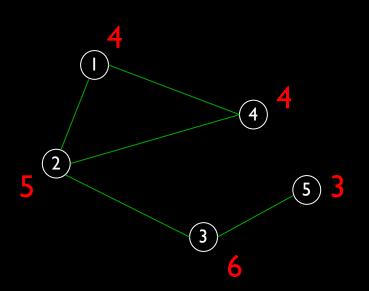
Context



Context



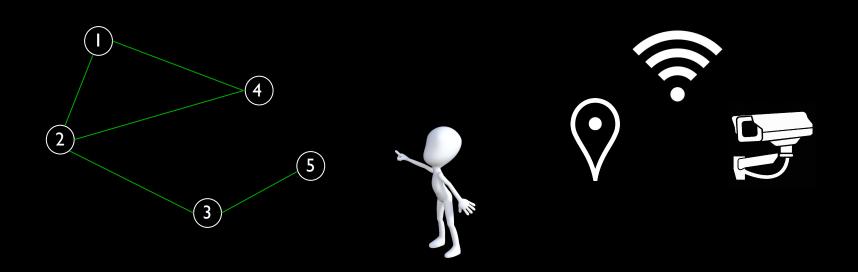
Context



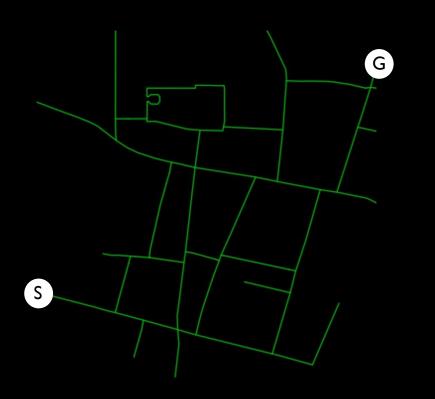
(Hillier '93) $\sum_{S=1}^{K} S \times N_{S}$

(2)
$$3 \times 1 + 1 \times 2 = 5$$
 (if $s = 2$)

Let's learn from human



Imitation Learning



Expert trajectory dataset

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

Reward Function

Maximum Entropy Inverse Reinfocement Learning

(Ziebart '08)

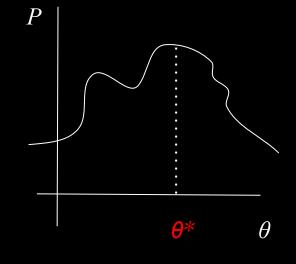
$$D = \{\zeta_{1}, \zeta_{2}, \zeta_{m}\}$$

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

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

$$Z = \oint \exp(R\theta(\zeta))$$

$$L(\theta)$$

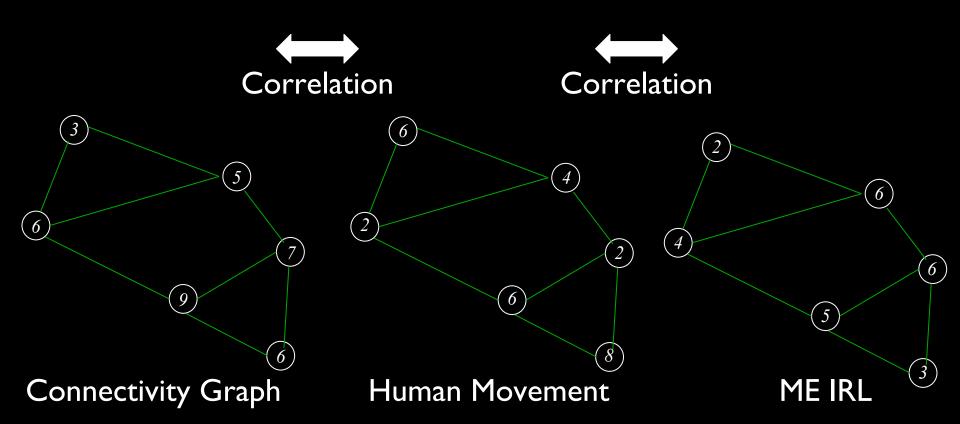


Maximum Entropy Inverse Reinfocement Learning

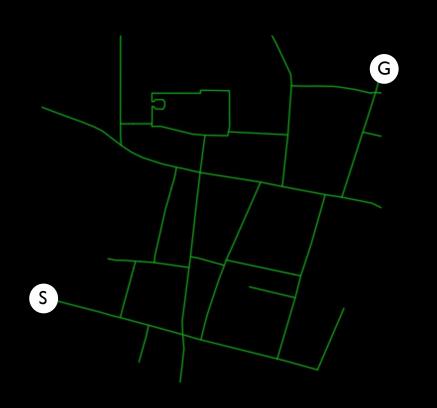
(Ziebart '08)

 $\theta_0 \leftarrow RewardParameterInitialization$ for i = 1 : N $R_i \leftarrow UpdateReward(\theta_i)$ Algorithm $\pi_i \leftarrow UpdatePolicy(R_i)$ $E \leftarrow PolicyPropagation(\pi_i)$ $\overline{Compute} \ \overline{\nabla L(\theta)}$ $\theta_0 \leftarrow UpdateRewardParameter(\nabla L(\theta))$

Evaluation



Problem that I need to overcome



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

Questions?