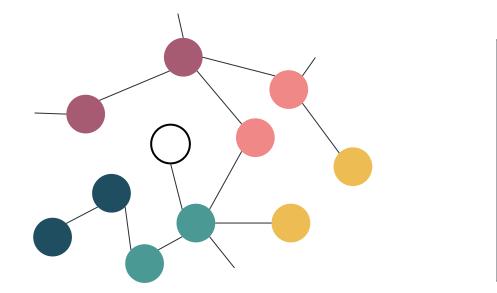
Directed Graph Embeddings in Pseudo-Riemannian Manifolds

Aaron Sim Maciej Wiatrak Angus Brayne Páidí Creed Saee Paliwal

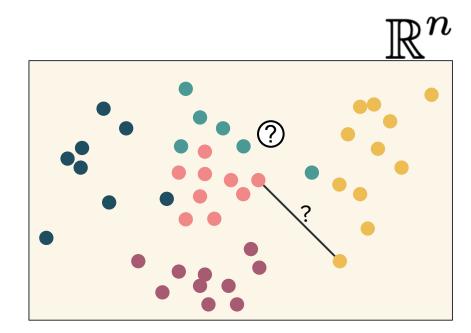


The Graph Embedding Problem:

Finding a vector representation for each node in a graph



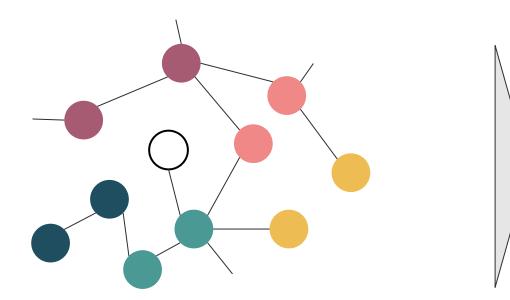
- Small social networks (**10² nodes**)
- Gene regulatory networks (**10⁵ nodes**)
- Knowledge graphs (**10⁷ nodes**)



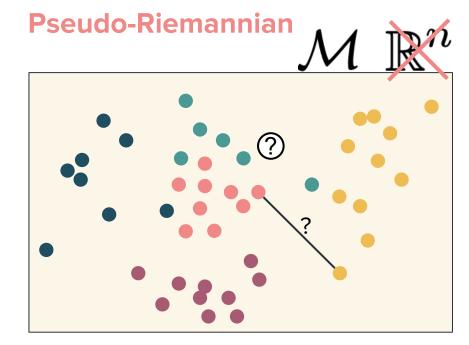
- Uncover hidden data features
- Node Classification, Link Prediction

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Finding a vector representation for each node in a graph

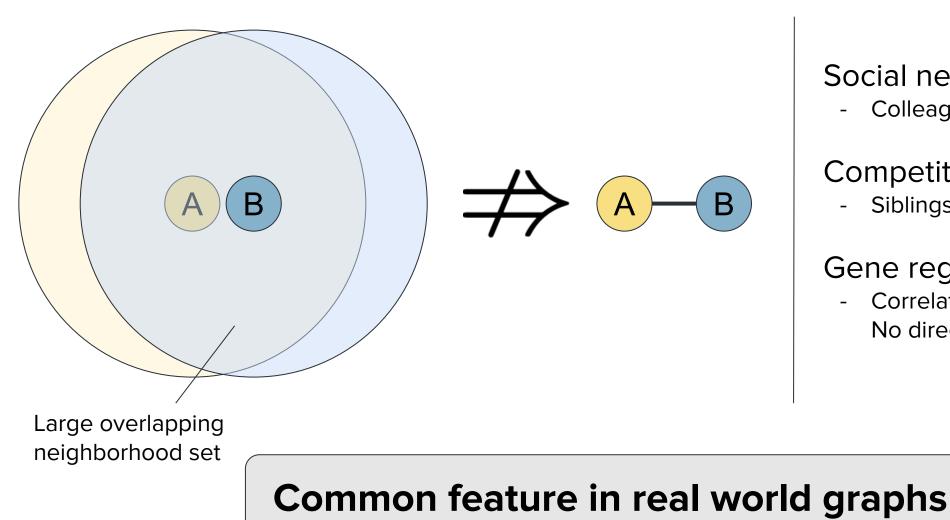


- Small social networks (**10² nodes**)
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- Uncover hidden data features
- Node Classification, Link Prediction

Semantic Similarity vs. Graph Edges



Social networks

Colleagues -

Competitive sports

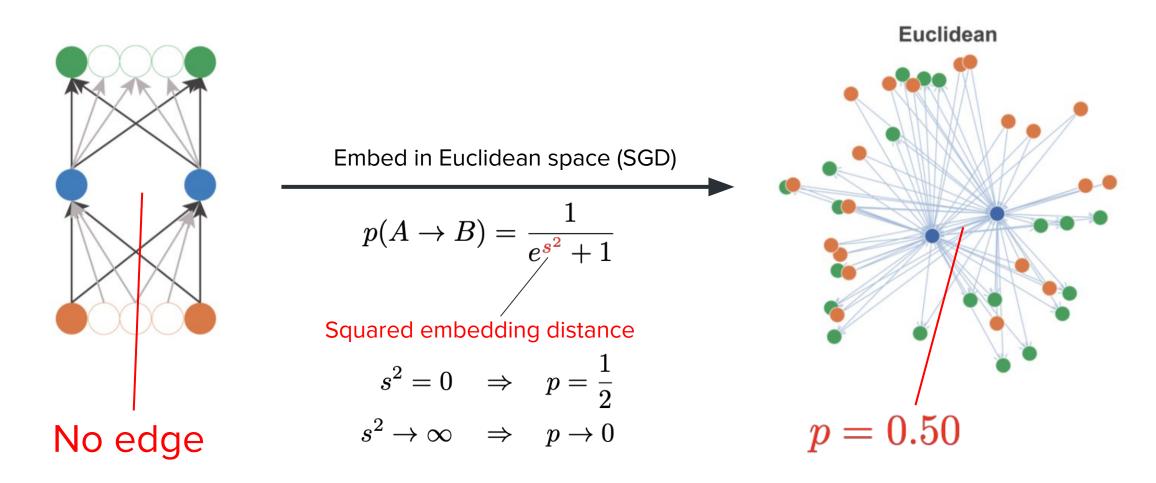
Siblings -

Gene regulatory networks

Correlated expression vs. -No direct causal link

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Riemannian manifolds cannot disambiguate semantic similarity and edges



Proposal 1: Embed in pseudo-Riemannian manifolds

No edge

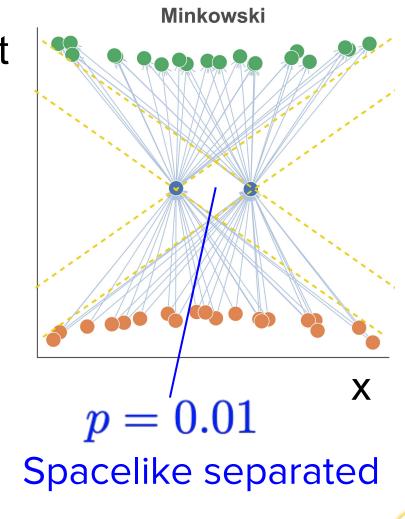
Embed in Minkowski spacetime (PR-SGD)

$$p(A \to B) = \frac{1}{e^{s^2} + 1}$$

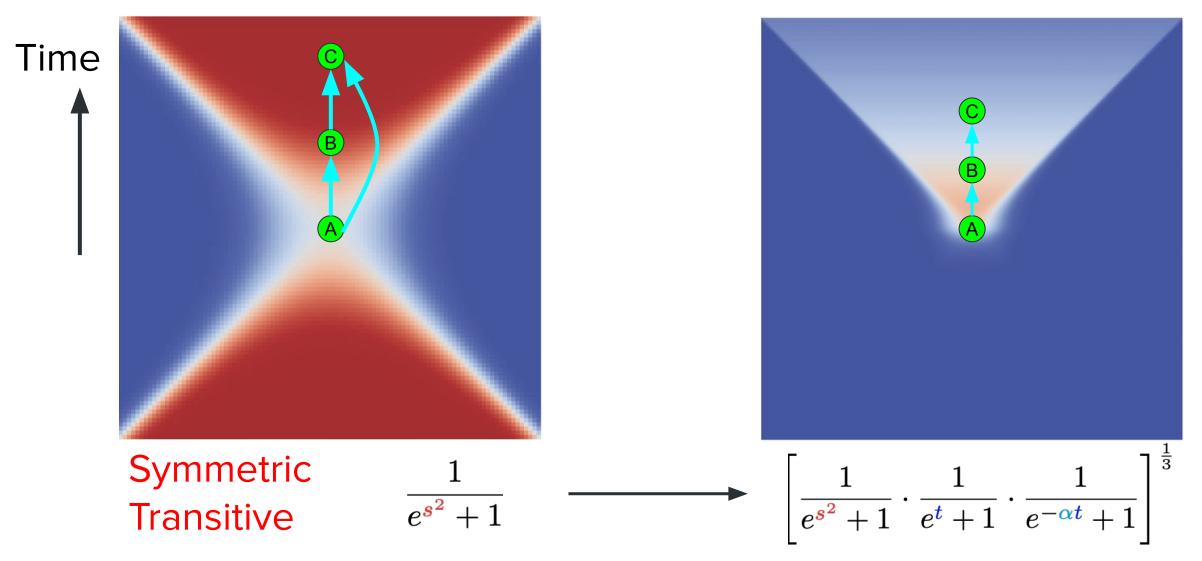
Squared embedding distance

$$s^{2} < 0 \quad \Rightarrow \quad \frac{1}{2} < p < 1$$
$$s^{2} = 0 \quad \Rightarrow \quad p = \frac{1}{2}$$
$$s^{2} \to \infty \quad \Rightarrow \quad p \to 0$$

$$s^2 = -(\Delta t)^2 + (\Delta x)^2$$



Proposal 2: *Triple Fermi-Dirac* likelihood function

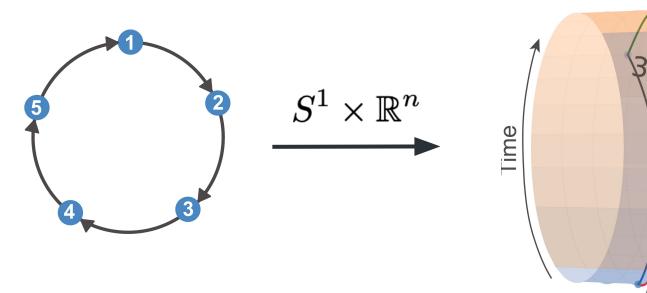


Benevolent^{AI} (7

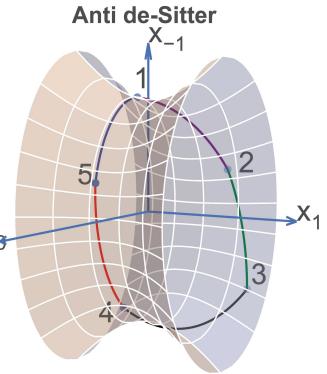
Proposal 3: Compact time coordinate for graph cycles

Cylindrical Minkowski

Space



3 1 4 X_G



Benevolent^{AI} (8

