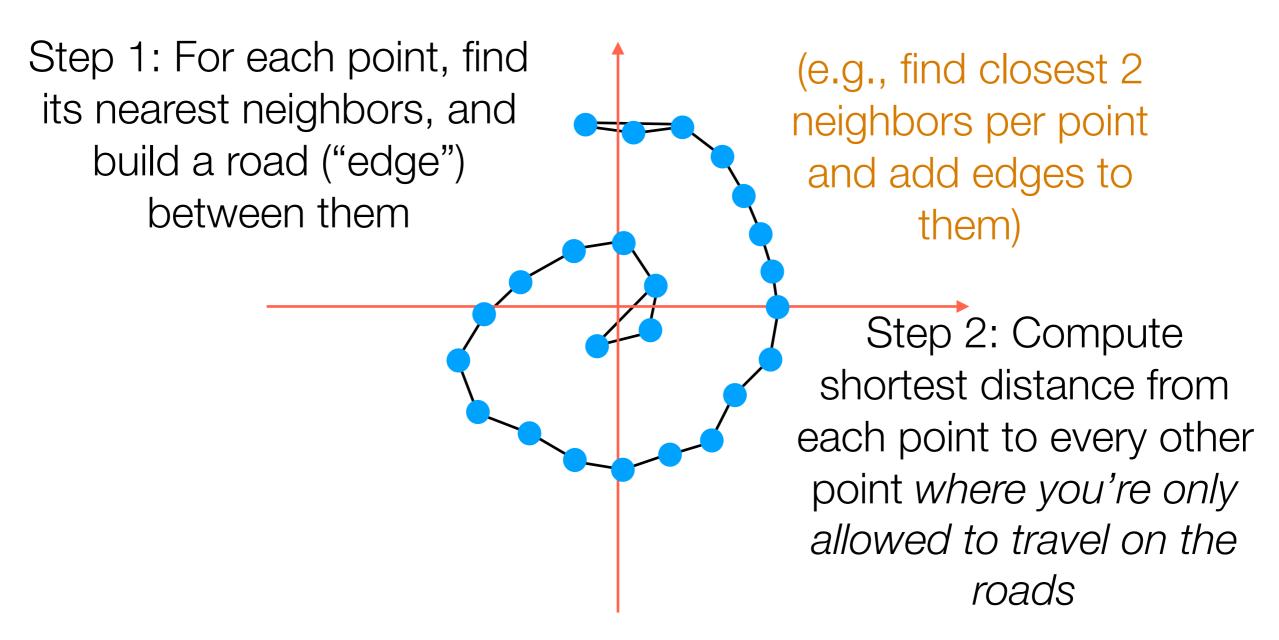


94-775/95-865 Lecture 4: Manifold learning

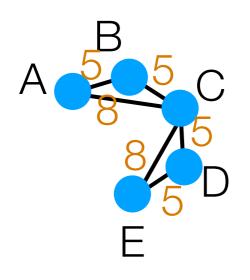
George Chen

Manifold Learning with Isomap



Step 3: It turns out that given all the distances between pairs of points, we can compute what the points should be (the algorithm for this is called *multidimensional scaling*)

In orange: road lengths



2 nearest neighbors of A: B, C

2 nearest neighbors of B: A, C

2 nearest neighbors of C: B, D

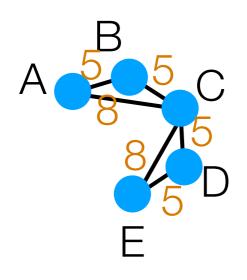
2 nearest neighbors of D: C, E

2 nearest neighbors of E: C, D

Build "symmetric 2-NN" graph (add edges for each point to its 2 nearest neighbors)

| • | А | В | С | D | Е |
|---|---|---|---|---|---|
| А | | | | | |
| В | | | | | |
| С | | | | | |
| D | | | | | |
| Е | | | | | |

In orange: road lengths



2 nearest neighbors of A: B, C

2 nearest neighbors of B: A, C

2 nearest neighbors of C: B, D

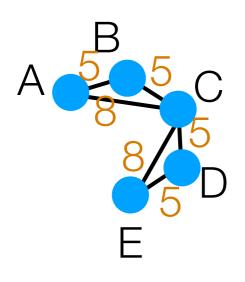
2 nearest neighbors of D: C, E

2 nearest neighbors of E: C, D

Build "symmetric 2-NN" graph (add edges for each point to its 2 nearest neighbors)

| | А | В | С | D | Е |
|---|---|---|---|---|---|
| А | 0 | | | | |
| В | | 0 | | | |
| С | | | 0 | | |
| D | | | | 0 | |
| Е | | | | | 0 |

In orange: road lengths



2 nearest neighbors of A: B, C

2 nearest neighbors of B: A, C

2 nearest neighbors of C: B, D

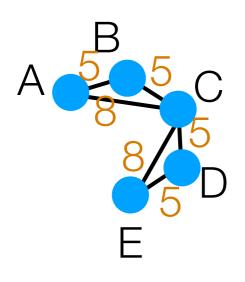
2 nearest neighbors of D: C, E

2 nearest neighbors of E: C, D

Build "symmetric 2-NN" graph (add edges for each point to its 2 nearest neighbors)

| | А | В | С | D | Е |
|---|---|---|---|---|---|
| А | 0 | 5 | | | |
| В | | 0 | 5 | | |
| С | | | 0 | 5 | |
| D | | | | 0 | 5 |
| Е | | | | | 0 |

In orange: road lengths



2 nearest neighbors of A: B, C

2 nearest neighbors of B: A, C

2 nearest neighbors of C: B, D

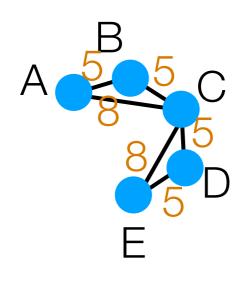
2 nearest neighbors of D: C, E

2 nearest neighbors of E: C, D

Build "symmetric 2-NN" graph (add edges for each point to its 2 nearest neighbors)

| | А | В | С | D | Е |
|---|---|---|---|---|---|
| А | 0 | 5 | 8 | | |
| В | | 0 | 5 | | |
| С | | | 0 | 5 | |
| D | | | | 0 | 5 |
| Е | | | | | 0 |

In orange: road lengths



2 nearest neighbors of A: B, C

2 nearest neighbors of B: A, C

2 nearest neighbors of C: B, D

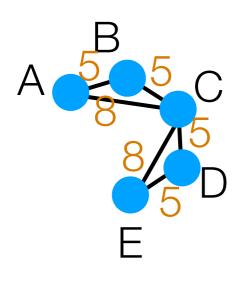
2 nearest neighbors of D: C, E

2 nearest neighbors of E: C, D

Build "symmetric 2-NN" graph (add edges for each point to its 2 nearest neighbors)

| | А | В | С | D | Е |
|---|---|---|---|----|---|
| А | 0 | 5 | 8 | 13 | |
| В | | 0 | 5 | | |
| С | | | 0 | 5 | |
| D | | | | 0 | 5 |
| Е | | | | | 0 |

In orange: road lengths



2 nearest neighbors of A: B, C

2 nearest neighbors of B: A, C

2 nearest neighbors of C: B, D

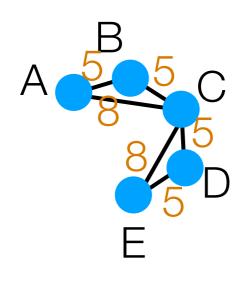
2 nearest neighbors of D: C, E

2 nearest neighbors of E: C, D

Build "symmetric 2-NN" graph (add edges for each point to its 2 nearest neighbors)

| | А | В | С | D | Е |
|---|---|---|---|----|----|
| А | 0 | 5 | 8 | 13 | 16 |
| В | | 0 | 5 | | |
| С | | | 0 | 5 | |
| D | | | | 0 | 5 |
| Е | | | | | 0 |

In orange: road lengths



2 nearest neighbors of A: B, C

2 nearest neighbors of B: A, C

2 nearest neighbors of C: B, D

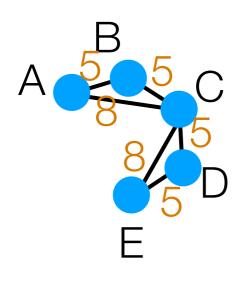
2 nearest neighbors of D: C, E

2 nearest neighbors of E: C, D

Build "symmetric 2-NN" graph (add edges for each point to its 2 nearest neighbors)

| | А | В | С | D | Е |
|---|---|---|---|----|----|
| А | 0 | 5 | 8 | 13 | 16 |
| В | | 0 | 5 | 10 | |
| С | | | 0 | 5 | |
| D | | | | 0 | 5 |
| Е | | | | | 0 |

In orange: road lengths



2 nearest neighbors of A: B, C

2 nearest neighbors of B: A, C

2 nearest neighbors of C: B, D

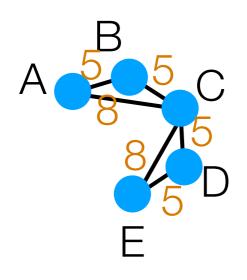
2 nearest neighbors of D: C, E

2 nearest neighbors of E: C, D

Build "symmetric 2-NN" graph (add edges for each point to its 2 nearest neighbors)

| | А | В | С | D | Е |
|---|---|---|---|----|----|
| А | 0 | 5 | 8 | 13 | 16 |
| В | | 0 | 5 | 10 | 13 |
| С | | | 0 | 5 | |
| D | | | | 0 | 5 |
| Е | | | | | 0 |

In orange: road lengths



2 nearest neighbors of A: B, C

2 nearest neighbors of B: A, C

2 nearest neighbors of C: B, D

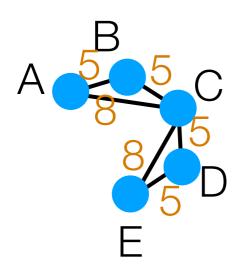
2 nearest neighbors of D: C, E

2 nearest neighbors of E: C, D

Build "symmetric 2-NN" graph (add edges for each point to its 2 nearest neighbors)

| | А | В | С | D | Е |
|---|---|---|---|----|----|
| А | 0 | 5 | 8 | 13 | 16 |
| В | | 0 | 5 | 10 | 13 |
| С | | | 0 | 5 | 8 |
| D | | | | 0 | 5 |
| Е | | | | | 0 |

In orange: road lengths



2 nearest neighbors of A: B, C

2 nearest neighbors of B: A, C

2 nearest neighbors of C: B, D

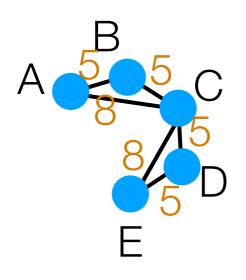
2 nearest neighbors of D: C, E

2 nearest neighbors of E: C, D

Build "symmetric 2-NN" graph (add edges for each point to its 2 nearest neighbors)

| | А | В | С | D | Е |
|---|----|----|---|----|----|
| А | 0 | 5 | 8 | 13 | 16 |
| В | 5 | 0 | 5 | 10 | 13 |
| С | 8 | 5 | 0 | 5 | 8 |
| D | 13 | 10 | 5 | 0 | 5 |
| Е | 16 | 13 | 8 | 5 | 0 |

In orange: road lengths



2 nearest neighbors of A: B, C

2 nearest neighbors of B: A, C

2 nearest neighbors of C: B, D

2 nearest neighbors of D: C, E

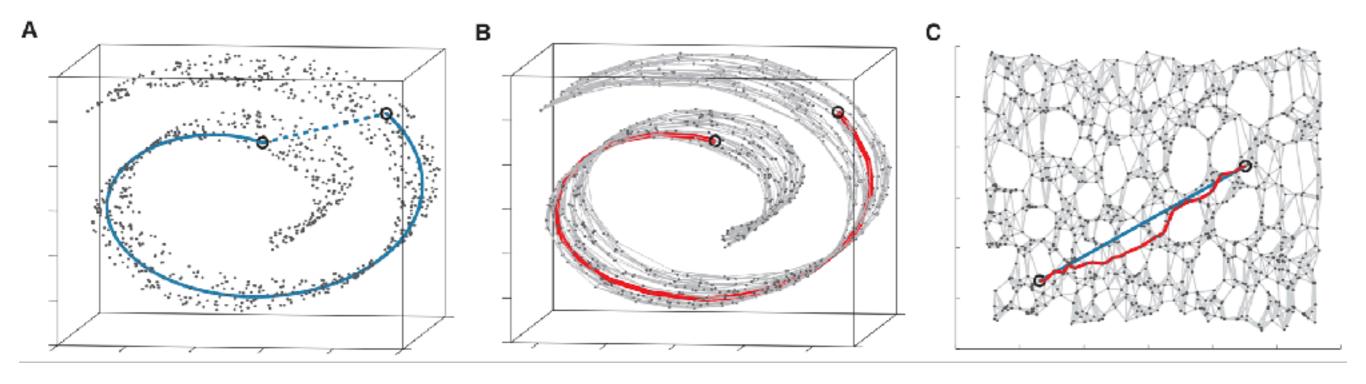
2 nearest neighbors of E: C, D

Build "symmetric 2-NN" graph (add edges for each point to its 2 nearest neighbors)

| | А | В | С | D | Е | | |
|---|---|----------|----------|----------|------------------|--|--|
| А | 0 | 5 | 8 | 13 | 16 | | |
| В | This matrix gets fed into multidimensional scaling to get | | | | | | |
| С | 81D | version | n of A, | B, C, D | , E ⁸ | | |
| D | ¹³ Th | e soluti | on is no | ot uniqu | Je! ⁵ | | |
| Е | 16 | 13 | 8 | 5 | 0 | | |

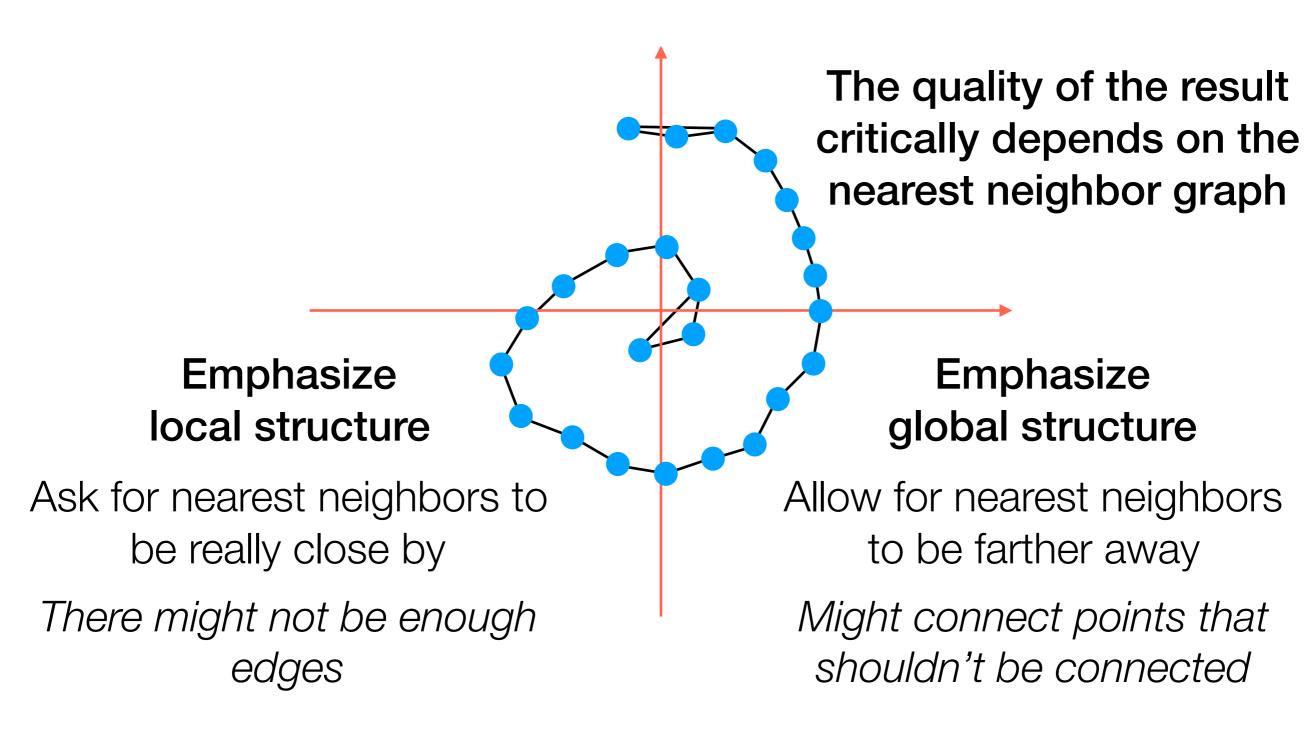
Demo

3D Swiss Roll Example



Joshua B. Tenenbaum, Vin de Silva, John C. Langford. A Global Geometric Framework for Nonlinear Dimensionality Reduction. Science 2000.

Some Observations on Isomap

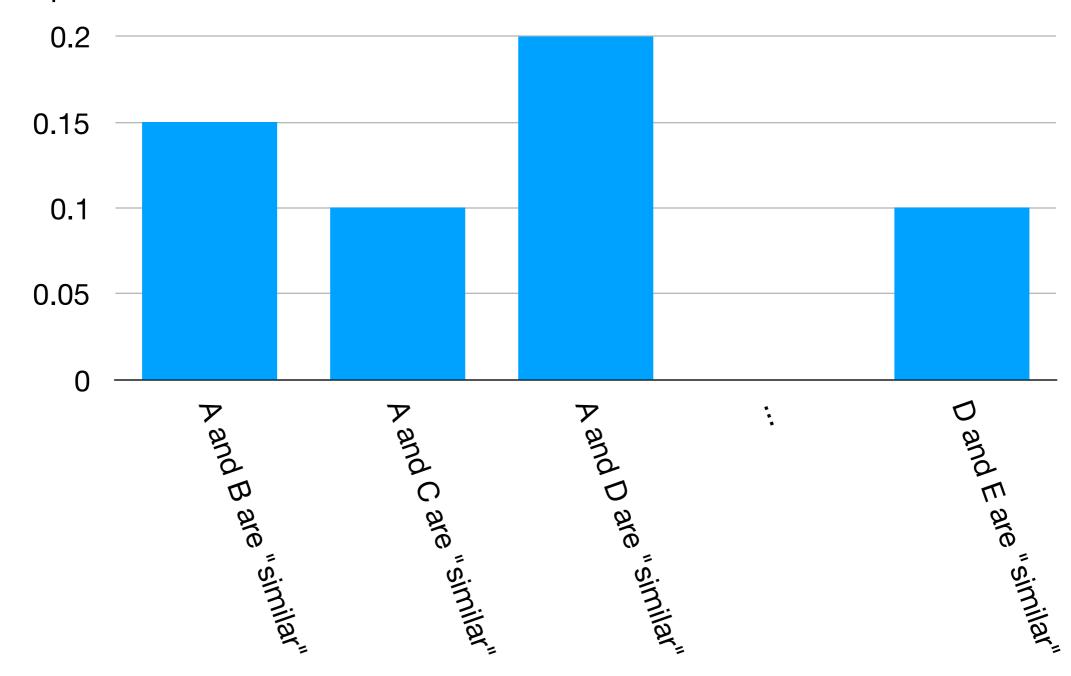


In general: try different parameters for nearest neighbor graph construction when using Isomap + visualize

t-SNE (t-distributed stochastic neighbor embedding)

t-SNE High-Level Idea #1

- Don't use deterministic definition of which points are neighbors
- Use probabilistic notation instead

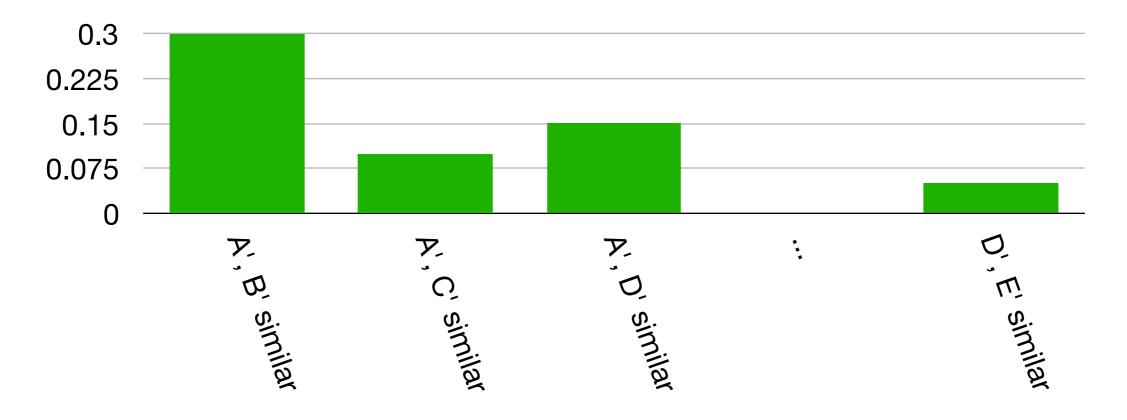


t-SNE High-Level Idea #2

 In low-dim. space (e.g., 1D), suppose we just randomly assigned coordinates as a candidate for a low-dimensional representation for A, B, C, D, E (I'll denote them with primes):



 With any such candidate choice, we can define a probability distribution for these <u>low-dimensional</u> points being similar

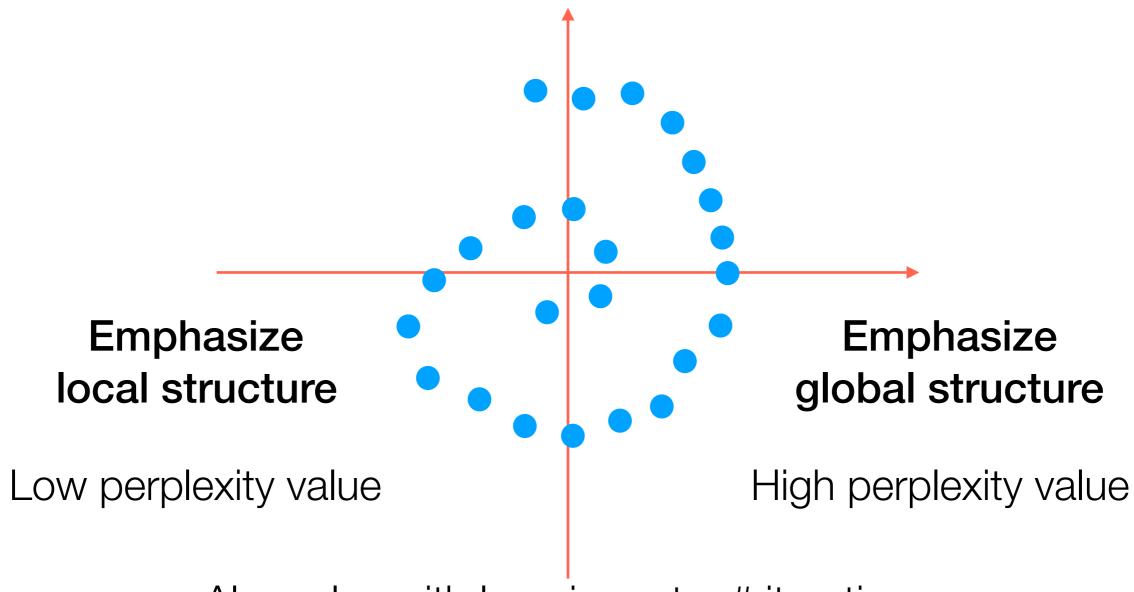


t-SNE High-Level Idea #3

 Keep improving low-dimensional representation to make the following two distributions look as closely alike as possible



t-SNE



Also: play with learning rate, # iterations
In practice, often people initialize with PCA
There are some other parameters (less critical)

Manifold Learning with t-SNE

Demo

t-SNE Interpretation

https://distill.pub/2016/misread-tsne/