

### **Unstructured Data Analysis**

### Lecture 5: Manifold learning

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# A Note on Design Choices

- Should I lowercase? Should I lemmatize? How do I count co-occurrences (at the sentence level? paragraph level? document level?), ... *lots of design choices!* 
  - When you do data analysis for a company/organization, often there is an **infinite number of design choices**
  - There usually will not be someone that tells you what is the "correct" way to choose all of these design choices
  - You have to make these decisions!
- If you're not sure about what to use, try multiple options and see for yourself how the output changes and whether this affects conclusions that are drawn from the analysis!
  - It's good for you to figure out what design choices lead to significant changes and what do not

PCA reorients data so axes explain variance in "decreasing order"
→ can "flatten" (*project*) data onto a few axes that captures most variance



Image source: http://4.bp.blogspot.com/-USQEgoh1jCU/VfncdNOETcI/AAAAAAAGp8/ Hea8UtE\_1c0/s1600/Blog%2B1%2BIMG\_1821.jpg



PCA would just flatten this thing and lose the information that the data actually lives on a 1D line that has been curved!



Image source: http://4.bp.blogspot.com/-USQEgoh1jCU/VfncdNOETcI/AAAAAAAGp8/ Hea8UtE\_1c0/s1600/Blog%2B1%2BIMG\_1821.jpg













This is the desired result

## Manifold Learning

- Nonlinear dimensionality reduction (in contrast to PCA which is linear)
- Find low-dimensional "manifold" that the data live on



Basic idea of a manifold:

1. Zoom in on any point (say, x)

2. The points near *x* look like they're in a lower-dimensional Euclidean space (e.g., a 2D plane in Swiss roll)

### Do Data Actually Live on Manifolds?



Image source: http://www.columbia.edu/~jwp2128/Images/faces.jpeg

### **Do Data Actually Live on Manifolds?**



Image source: http://www.adityathakker.com/wp-content/uploads/2017/06/wordembeddings-994x675.png

### **Do Data Actually Live on Manifolds?**



Mnih, Volodymyr, et al. Human-level control through deep reinforcement learning. Nature 2015.

# Manifold Learning with Isomap



Step 3: It turns out that given all the distances between pairs of points, we can compute what the low-dimensional 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	E
А					
В					
С					
D					
E					



#### 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	E
А	0				
В		0			
С			0		
D				0	
E					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	E
А	0	5			
В		0	5		
С			0	5	
D				0	5
E					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	E
А	0	5	8		
В		0	5		
С			0	5	
D				0	5
E					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	E
А	0	5	8	13	
В		0	5		
С			0	5	
D				0	5
E					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	E
А	0	5	8	13	16
В		0	5		
С			0	5	
D				0	5
E					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	E
А	0	5	8	13	16
В		0	5	10	
С			0	5	
D				0	5
E					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	E
А	0	5	8	13	16
В		0	5	10	13
С			0	5	
D				0	5
E					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	E
А	0	5	8	13	16
В		0	5	10	13
С			0	5	8
D				0	5
E					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)

	A	В	С	D	E
А	0	5	8	13	16
В	5	0	5	10	13
С	8	5	0	5	8
D	13	10	5	0	5
E	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	E
А	0	5	8	13	16
В	T multi	his mat <i>dimen</i> s	rix gets <i>ional</i> so	s fed int c <i>aling</i> to	o get
С	81D	versior	n of A,	B, C, D	, Е <sup>8</sup>
D	<sup>13</sup> Th	e soluti	on is no	ot uniqu	Je! 5
E	16	13	8	5	0



### Isomap



Demo

### **3D Swiss Roll Example**

Key idea: true distance on manifold is the blue line



We're approximating the blue line with the red line (poor choice of # nearest neighbors can make approximation bad)

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