Machine learning algorithms

PCA, #2

2020-10-5

CSCI 471 / 571, Fall 2020 Kameron Decker Harris

Recommended talk!

- Nathan Kutz, 3 p.m. on Thursday
- Western Washington Data-driven Discovery Seminar Series
- Machine Learning for Science: Data-Driven Discovery Methods for Governing equations, Coordinates and Sensors

Homework questions?

```
total # = Ntest
                       16,64,144
                       4x4, 8×8, 12×12
                 for x, in np. linspace (0,1,4)
Problem 11
                          vec_x = up.array ([x, x])
  f(\vec{x})
      true_fun (x)
                                     pred-wrap (x) = lambda x:
pred-fun(x, beta)
      pred-fun (x, beta)
                       leugte es degree
```

Problem 6 ||XB-g||2 = ~TA ~ + VTB~+ (C) will depend on one of the least square terms $X, \overline{B}, \overline{y}$ 11711 = inner product ... rules of transposes, etc. Problem 10 $\overrightarrow{X} = \begin{pmatrix} a \\ b \end{pmatrix} \qquad \begin{bmatrix} a & b \end{bmatrix} \xrightarrow{\rho=3} \qquad \begin{bmatrix} 1 & a & b & a^2 & b^2 \\ & & & & \\ & & &$ Problem 4

Problem 7(7.1) write in $f(\vec{u}, \vec{v})$ output F

problem 4

Problem 7(7.1) write in $f(\vec{u}, \vec{v})$ use # u to get gradient $A \in \mathbb{R}^{m \times n}$ Same as "A is $m \times n$ matrix" $M\vec{\beta} = \vec{c}$ $\vec{x} \in \mathbb{R}^n$ " \vec{x} is n-dim sector (column)"

P(A, SVD) Explained variance

$$X = U \int V^{T}$$
 $\sigma_{i}^{2} = variance captured by$
 $singular values, larget = smallest$
 $singular values, larget = smallest$

$$\frac{\sum_{i=1}^{3} \sigma_{i}^{2}}{\sum_{j=1}^{cank(x)} \sigma_{j}^{2}} = 86\% \quad \text{Variance explained}$$

$$= 86\% \quad \text{by top 3 components}$$

$$= \frac{1}{\sqrt{3}} = \frac{1}{\sqrt{3}}$$

Dirensionality reduction to
$$r$$
 dimensions $< d$.

 $V_r = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$
 $\forall v \in \mathbb{R}^d$
 $\forall v \in \mathbb{R}^d$
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$$X = \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \end{bmatrix}, x = \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \end{bmatrix}$$

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$$X = \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \end{bmatrix}, x = \begin{bmatrix} \frac{1}{2} \\ \frac$$

$$(A)(A \times r) \qquad (A \times r) \qquad ($$

Examples

Super popular

- 3.6 M
- Search "PCA", "principal components" on Scholar
- Credit for some of slides: Bing Brunton

The Shapes of Stories by Kurt Vonnegut

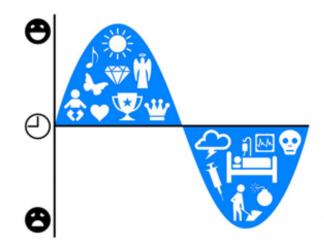
T Turt Vonnegut gained worldwide fame and adoration through the publication of his novels, including Slaughterhouse-Five, Cat's Cradle, Breakfast of Champions, and more.

But it was his rejected master's thesis in anthropology that he called his prettiest contribution to his culture.

The basic idea of his thesis was that a story's main character has ups and downs that can be graphed to reveal the story's shape.

The shape of a society's stories, he said, is at least as interesting as the shape of its pots or spearheads. Let's have a look.

Designer: Maya Eilam, www.mayaeilam.com Sources: A Man without a Country and Palm Sunday by Kurt Vonnegut



Man in Hole

Boy Meets Girl

From Bad to Worse



The main character gets into trouble
The main character comes across then gets out of it again and ends up something wonderful, gets it, loses better off for the experience.



Arsenic and Old Lace Harold & Kumar Go To



it, then gets it back forever.



Jane Eyre



Eternal Sunshine of the Spotless Mind



The main character starts off poorly then gets continually worse with no hope for improvement.



The Metamorphosis



The Twilight Zone



The story has a lifelike ambiguity that keeps us from knowing if new developments are good or bad.



Hamlet

The Sopranos

Creation Story

Old Testament

New Testament

Cinderella



In many cultures' creation stories, humankind receives incremental gifts from a deity. First major staples like the earth and sky, then smaller things like sparrows and cell phones. Not a common shape for Western stories, however.



Humankind receives incremental gifts from a deity, but is suddenly ousted from good standing in a fall



Humankind receives incremental gifts from a deity, is suddenly ousted from good standing, but of enormous proportions.



Great Expectations with Dickens' alternate ending

then receives off-the-charts bliss.



It was the similarity between the shapes of Cinderella and the New Testament that thrilled Vonnegut for the first time in 1947 and then over the course of his life as he continued to write essays and give

http://mayaeilam.com/2012/01/01/the-shapes-of-stories-a-kurt-vonnegut-infographic/

Man in Hole

Boy Meets Girl

From Bad to Worse

Which Way Is Up?



The main character gets into trouble then gets out of it again and ends up better off for the experience.



Arsenic and Old Lace



Harold & Kumar Go To White Castle



The main character comes across something wonderful, gets it, loses it, then gets it back forever.



Jane Eyre



Eternal Sunshine of the Spotless Mind



The main character starts off poorly then gets continually worse with no hope for improvement.



The Metamorphosis



The Twilight Zone



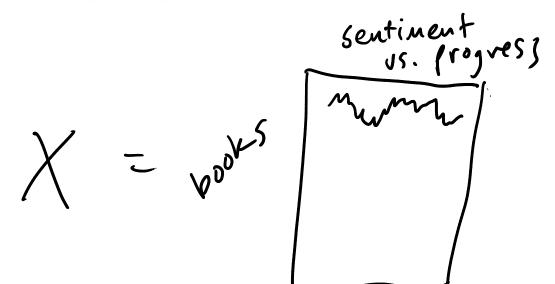
The story has a lifelike ambiguity that keeps us from knowing if new developments are good or bad.

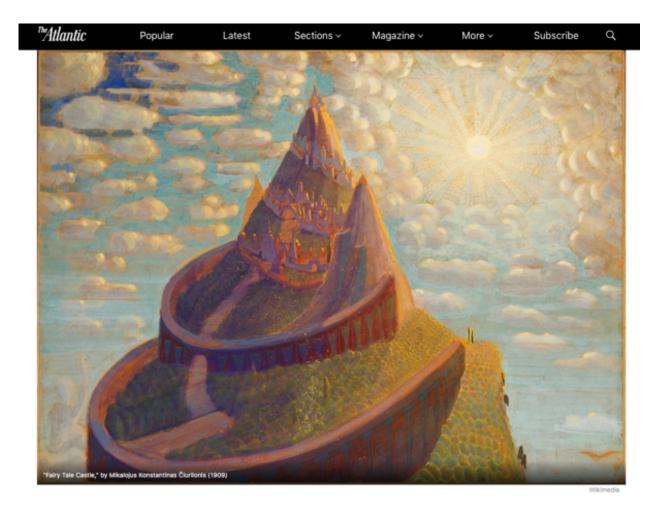


Hamlet



The Sopranos





The Six Main Arcs in Storytelling, as Identified by an A.I.

A machine mapped the most frequently used emotional trajectories in fiction, and compared them with the ones readers like best.

ADRIENNE LAFRANCE | JUL 12, 2016 | TECHNOLOGY

https://www.theatlantic.com/technology/archive/ 2016/07/the-six-main-arcs-in-storytelling-identified-bya-computer/490733/

The emotional arcs of stories are dominated by six basic shapes

Andrew J Reagan III, Lewis Mitchell, Dilan Kiley, Christopher M Danforth and Peter Sheridan Dodds

EPJ Data Science 2016 5:31 DOI: 10.1140/epjds/s13688-016-0093-1 © Reagan et al. 2016

Received: 2 July 2016 | Accepted: 26 October 2016 | Published: 4 November 2016

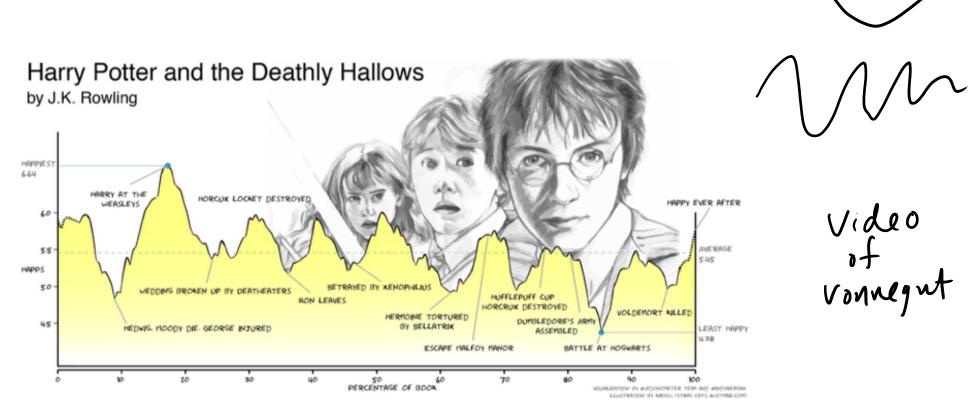


Fig 2

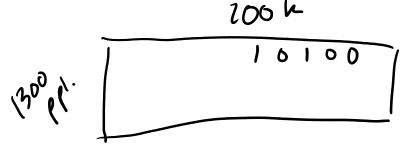
LETTERS

Genes mirror geography within Europe

John Novembre^{1,2}, Toby Johnson^{4,5,6}, Katarzyna Bryc⁷, Zoltán Kutalik^{4,6}, Adam R. Boyko⁷, Adam Auton⁷, Amit Indap⁷, Karen S. King⁸, Sven Bergmann^{4,6}, Matthew R. Nelson⁸, Matthew Stephens^{2,3} & Carlos D. Bustamante⁷

3,192 European individuals, PORES project each genotyped at 500,568 SNP loci

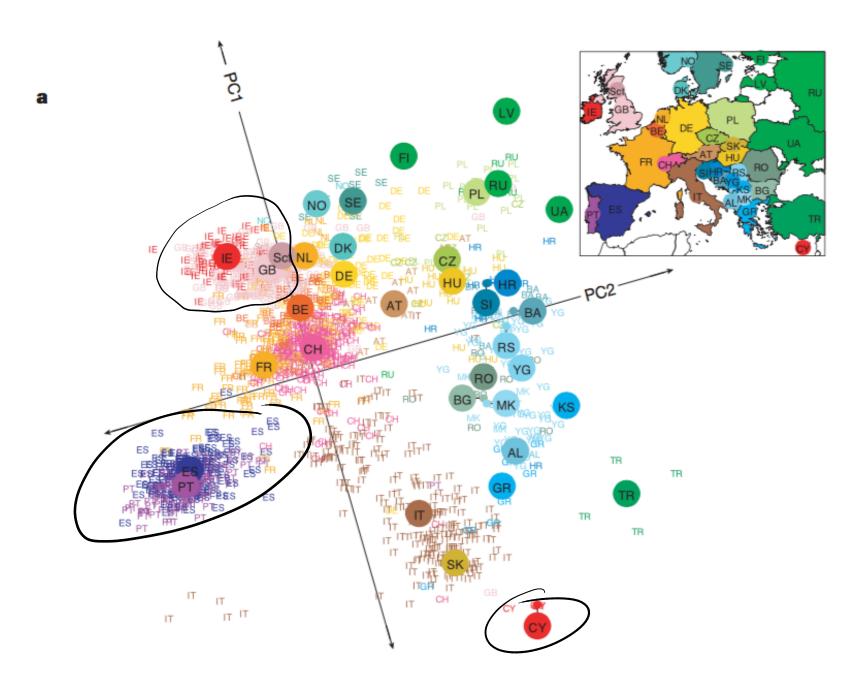
after stringent selection criteria, 1,387 individuals 197,145 loci

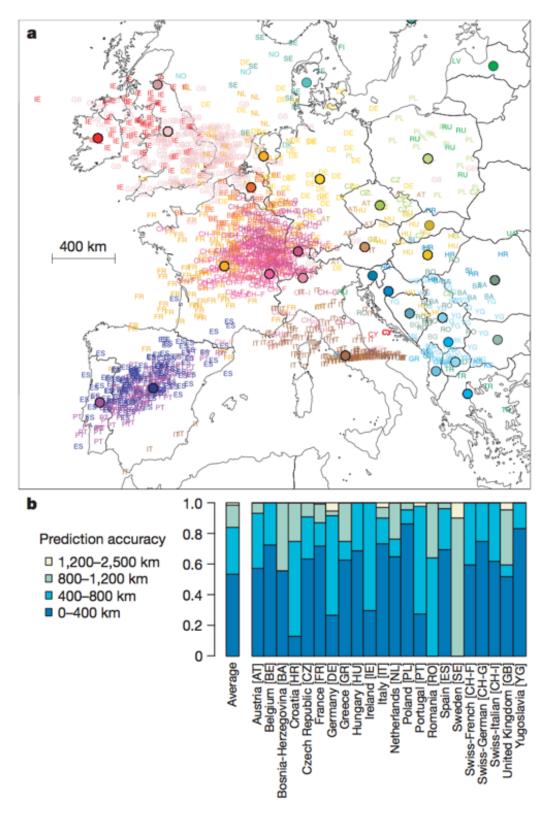


m = 197,145 measurements

Data Matrix

that's a huge amount of high-dimensional data! we can't even look at that sensibly...





Cross-validated Accuracy



https://github.com/NovembreLab/Novembre_etal_2008_misc

E-ARTICLE

Radiating despite a Lack of Character: Ecological Divergence among Closely Related, Morphologically Similar Honeyeaters (Aves: Meliphagidae) Co-occurring in Arid Australian Environments

Eliot T. Miller, 1,2,3,4,* Sarah K. Wagner, 4,5 Luke J. Harmon, 3,6 and Robert E. Ricklefs1

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Submitted December 11, 2015; Accepted September 14, 2016; Electronically published December 20, 2016 Online enhancements: appendix, supplemental PDF. Dryad data: http://dx.doi.org/10.5061/dryad.r82jg.

