

# JL Lemma - Extensions & Applications

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# Agenda

- 1 JL Lemma - Reminder
- 2 Database Friendly Random Projections
- 3 Compressed Least Squares Regression
- 4 Application - Modelling Computer Prices
- 5 Application - Visualizing Countries' Development

## Formal Statement

Let  $\epsilon \in (0, \frac{1}{2})$ . Let  $Q \subset \mathbb{R}^d$  be a set of  $n$  points and  $k = \frac{20 \log n}{\epsilon^2}$ . There exists a Lipschitz mapping  $f : \mathbb{R}^d \rightarrow \mathbb{R}^k$  such that for all  $u, v \in Q$

$$(1 - \epsilon) \|u - v\|^2 \leq \|f(u) - f(v)\|^2 \leq (1 + \epsilon) \|u - v\|^2$$

## What it Means

If we choose a matrix  $A$  where the entries of  $A$  are  $\mathcal{N}(0, 1)$  then for vectors  $u, v \in \mathbb{R}^d$  then  $\|u - v\|^2 \approx \|Au - Av\|^2$

# Problem!

- For  $x_1, \dots, x_n \in \mathbb{R}^d$ , and projection matrix  $A \in \mathbb{R}^{d \times k}$  where  $k < d$ , matrix multiplication  $O(ndk)$ .
- SQL can easily generate Uniform  $[0, 1]$  random variables, harder to generate normal random variables

## Generate Uniform $[0, 1]$ Random Variable

```
select rand()
```

## Generate Uniform $\{-1, 1\}$ Random Variable

```
select ceiling(rand()-.5)*2-1
```

## Generate Normal Random Variable

```
select randNum1 = rand()
```

```
select randNum2 = rand()
```

```
select value1 = round( (sqrt( -2.0*log(randNum1) )  
    *cos (2pi*randNum2) )*stdDev, precision)+ mean
```

```
select value2 = round((sqrt(-2.0*log(randNum1))  
    *sin(2pi*randNum2))*stdDev, precision)+ mean
```

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  - Moments
  - Bound on  $\chi^2$  random variable
- How about we make  $A$  with nicer random variables!

# Database Friendly Random Projections - Achiloptas (2001)

How about we make  $A$  with nicer random variables! Let  $A$ 's entries  $\{a_{ij}\}_{1 \leq i \leq d, 1 \leq j \leq k}$  be as follows

$$a_{ij} = \begin{cases} 1 & \text{with probability } \frac{1}{2} \\ -1 & \text{with probability } \frac{1}{2} \end{cases} \quad (1)$$

or

$$a_{ij} = \begin{cases} \sqrt{3} & \text{with probability } \frac{1}{6} \\ 0 & \text{with probability } \frac{2}{3} \\ -\sqrt{3} & \text{with probability } \frac{1}{6}. \end{cases} \quad (2)$$

- For random variable 1 we only generate random variables  $\in \{-1, +1\}$ !
- For random variable 2 we can throw away 2/3 of the data!

Let  $Q \subset \mathbb{R}^d$  be a set of  $n$  points. For a given  $\epsilon, \beta > 0$  let

$$k_0 = \frac{4 + 2\beta}{\epsilon^2/2 - \epsilon^3/3} \log n$$

and for some integer  $k \geq k_0$  let  $A \in \mathbb{R}^{d \times k}$  be a random matrix where its entries are independent random variables from the prveiously mentioned distributions. Let

$$E = \frac{1}{\sqrt{k}}QA$$

and  $f : \mathbb{R}^d \rightarrow \mathbb{R}^k$  map the  $i^{th}$  row of  $Q$  to the  $i^{th}$  row of  $E$ . Then with probability of at least  $1 - n^{-\beta}$  for all  $u, v \in Q$

$$(1 - \epsilon)||u - v||^2 \leq ||f(u) - f(v)||^2 \leq (1 + \epsilon)||u - v||^2.$$

**Moral of the story: JL Lemma still works!**

**Classic Regression**

$\{x_j, y_j\}_{1 \leq j \leq n}$ ,  $x_j \in \mathbb{R}^d$ ,  $y_j \in \mathbb{R}$ , want to find  $\hat{\beta} \in \mathbb{R}^{d+1}$  such that

$$\hat{\beta} = \arg \min_{(\beta_0, \dots, \beta_d)} \sum_{j=1}^n (y_j - (\beta_0 + \beta_1 x_{j,1} + \dots + \beta_d x_{j,d}))^2$$

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**Problem!**

Suppose  $d > n$ . Where  $X \in \mathbb{R}^{n \times d}$  is a matrix of the  $x$ 's stacked. To solve for  $\hat{\beta}$  we use

$$\hat{\beta} = (X^T X)^{-1} X^T Y.$$

But if  $X^T X$  invertible  $\Rightarrow X$  has linearly independent columns. But  $X$  has more columns than rows, therefore this is impossible!

Leads to solution not being unique.

**Solution**

Multiply  $X$  by a random matrix to reduce its dimensionality.

For example if  $X \in \mathbb{R}^{100 \times 10000}$ ,  $A \in \mathbb{R}^{10000 \times 10} \Rightarrow XA \in \mathbb{R}^{100 \times 10}$ .

$$\hat{\beta} = \arg \min_{(\beta_0, \dots, \beta_{10})} \sum_{j=1}^n (y_j - (\beta_0 + \beta_1[XA]_{j,1} + \dots + \beta_d[XA]_{j,d}))^2$$

A tractable problem!

## Bound on Squared Loss of Reduced Function (Remark 1)

$$\underbrace{\mathbb{E}[\|\hat{g}_L - f^*\|_P^2]}_{\text{Expected squared error of reduced dimension function}} = O\left(\|\alpha^+\| \sqrt{\mathbb{E}[\|\varphi(X)\|^2]} \frac{\log n/\delta}{\sqrt{n}} + \underbrace{\inf_{f \in \mathcal{F}_d} \|f - f^*\|_P^2}_{\text{Expected squared error of ideal function, without reducing dimension}}\right)$$

So even if we are doing regression in this reduced dimension setting, as  $n$  grows we can still achieve something close to optimal.



## Problem

Predict sales of 231 tablet computers each week for 24 weeks period ( $n = 5544$ ).

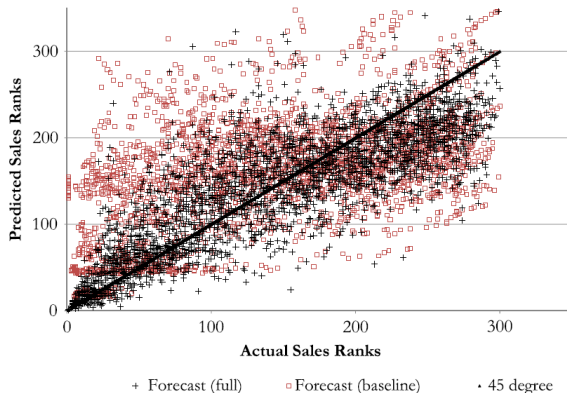
## Features

- 14 dimesions: features that are characteristics of each tablet (e.g. RAM, storage, weight)
- 20,068 dimensions: Bag of words - word counts from customer reviews of each product

## Solution

Using a random projection, reduce dimension of word vector to  $k = 300$ .

	Baseline Model (no words)	Full Model (with words)
$R^2$	34.2%	86.2%
Forecast Error	163.5%	37.2 %



The forecasts of sales ranks from the full model are closer to actual sales ranks than the baseline model. One can also see the baseline model produces quite a few forecasts which are wildly off.

## Question

Can we use random projections to better visualize high dimensional data.

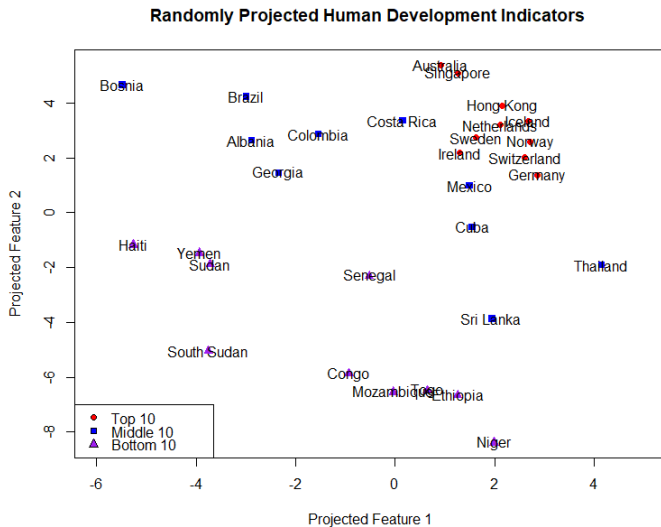
## Dataset

- Human Development Indicators from the United Nations for Top 10, Middle 10, Bottom 10 based on Human Development Index
- For each country retrieve most recent data on: dependency ratio, proportion urban population, mean years of school completed, CO<sup>2</sup> per capita, life expectancy, proportion internet users, proportion with basic drinking water, and unemployment rate.
- Normalized Data & Removed NAs

## Solution

Randomly project this dataset ( $n = 30$ ,  $p = 8$ ) to two dimensions and plot.

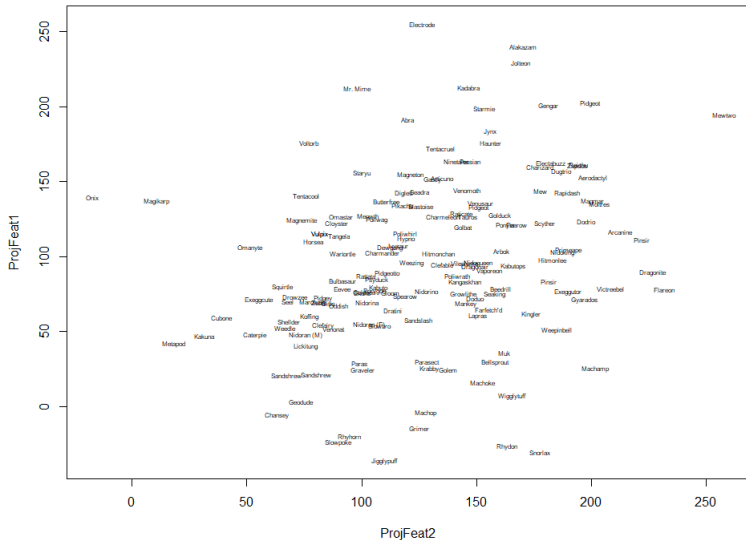
# Application - Visualizing Countries' Development



Thank You!

## Just for Fun - Visualizing 150 Pokemon Stats (All)

## Projected Pokemon



# Just for Fun - Visualizing 150 Pokemon Stats (Select)

