

*Michael Grossberg*

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# Intro to Data Science

## CS59969

PCA/SVD Dimension  
Reduction

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# Dimension Reduction (Linear)



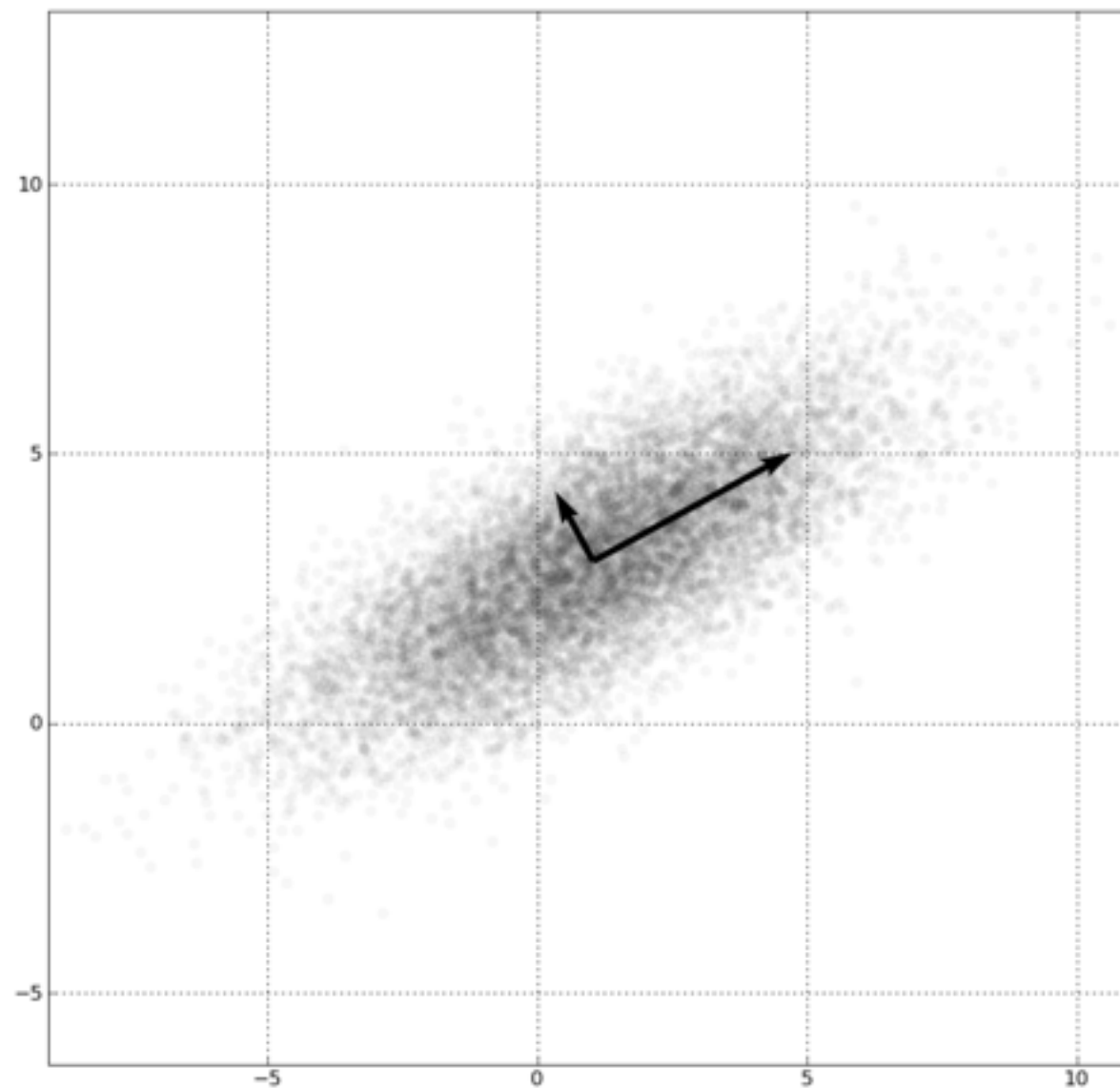
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# Principal Component Analysis (PCA)

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# 1-D mean, stdev

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Mean

$$\mu = E[x] = \frac{1}{n} \sum_{i=0}^n x_i$$

Variance = (Standard Deviation)<sup>2</sup>

$$\sigma^2 = E[(x - \mu)^2] = \frac{1}{n} \sum_{i=0}^n (x_i - \mu)^2$$

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# Normal (1-D ) distribution

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$$p(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{(x - \mu)^2}{2\sigma^2}}$$



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# N-D mean, (co-)variance

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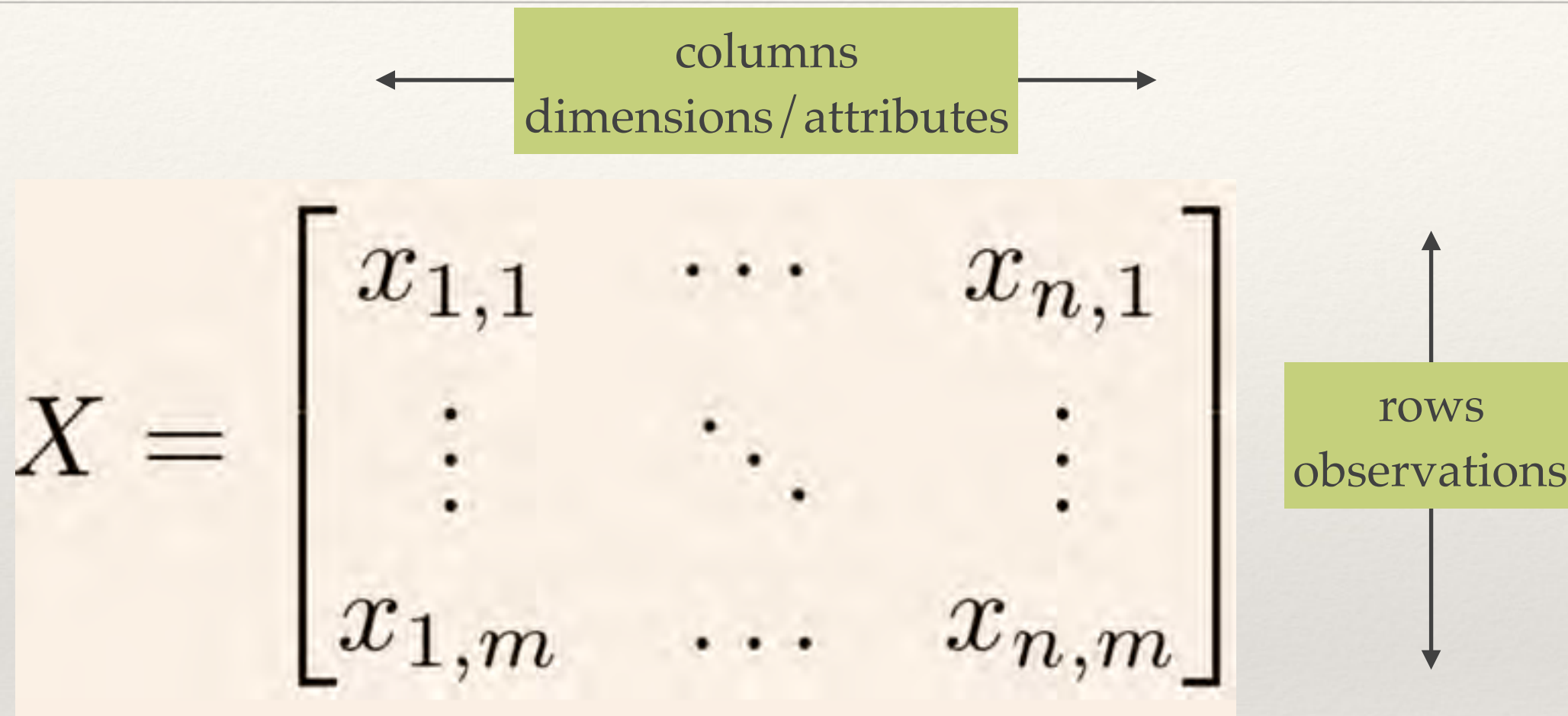
N-D Mean

$$\mu = E[\mathbf{x}] = \frac{1}{n} \sum_{i=0}^n \mathbf{x}_i = \left\{ \frac{1}{n} \sum_{k=0}^n x_{i,k} \right\}$$

N-D Covariance

$$\Sigma_{i,j} = \frac{1}{n} \sum_{k=0}^n (x_{i,k} - \mu)(x_{j,k} - \mu)$$

# Matrix Versions



A diagram illustrating the structure of a matrix  $X$ . The matrix is shown as a large square with elements  $x_{1,1}$ ,  $\dots$ ,  $x_{n,1}$  in the first row,  $\vdots$  in the middle, and  $x_{1,m}$ ,  $\dots$ ,  $x_{n,m}$  in the last row. Above the matrix, a horizontal double-headed arrow points to a green box containing the text "columns" and "dimensions / attributes". To the right of the matrix, a vertical double-headed arrow points to a green box containing the text "rows" and "observations".

$$X = \begin{bmatrix} x_{1,1} & \dots & x_{n,1} \\ \vdots & \ddots & \vdots \\ x_{1,m} & \dots & x_{n,m} \end{bmatrix}$$

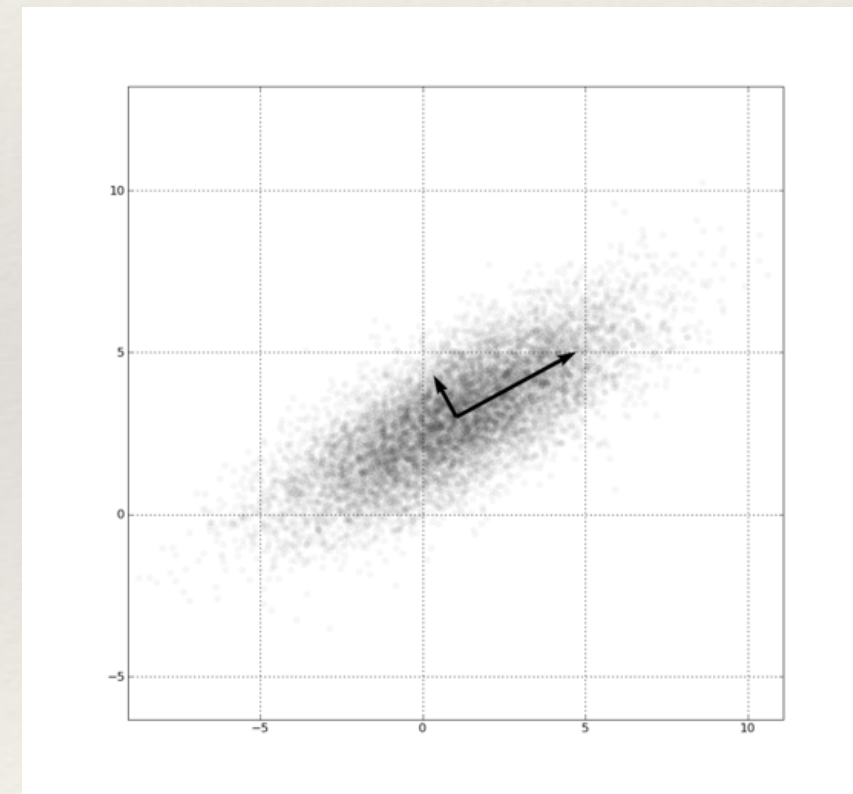
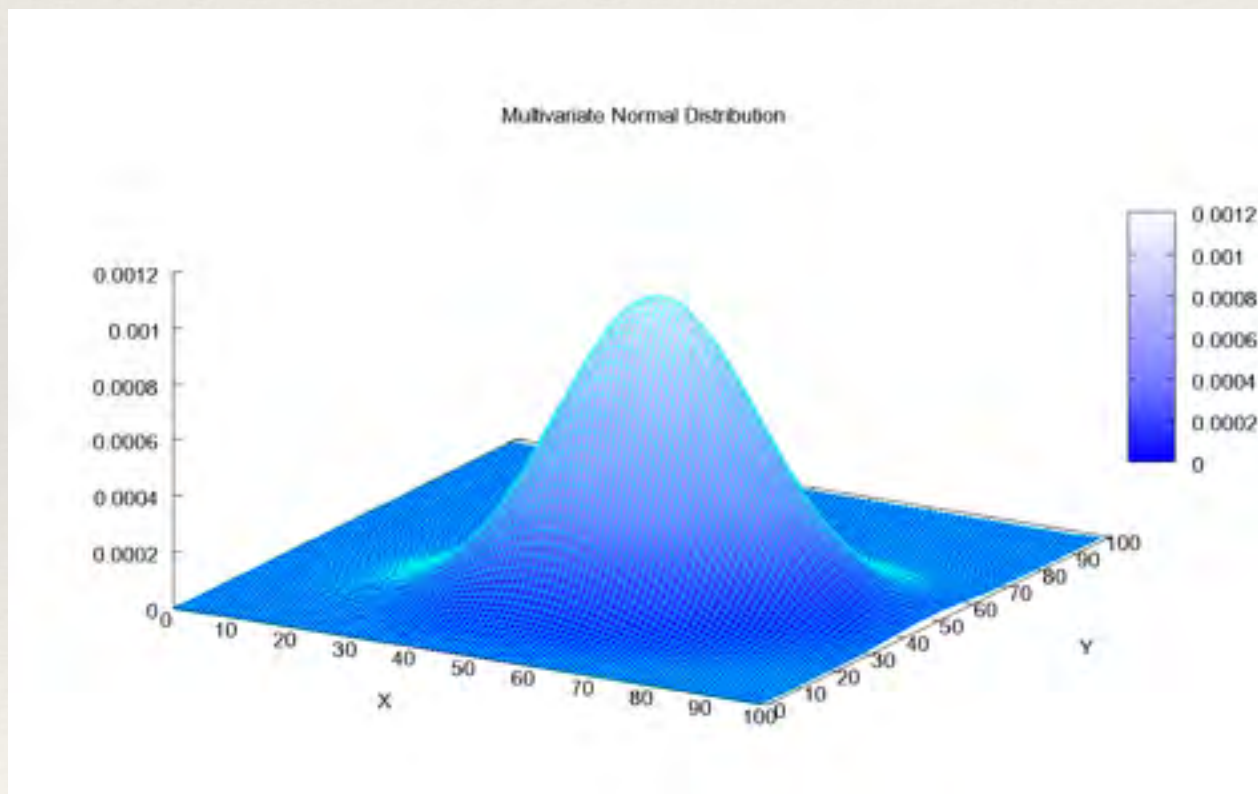
N-D Covariance

$$\Sigma = \check{X}^T \check{X}$$



# N-D Normal Distribution

$$f_{\mathbf{x}}(x_1, \dots, x_k) = \frac{1}{\sqrt{(2\pi)^k |\Sigma|}} \exp \left( -\frac{1}{2} (\mathbf{x} - \boldsymbol{\mu})^T \Sigma^{-1} (\mathbf{x} - \boldsymbol{\mu}) \right),$$



# Diagonalization

 $\Sigma$ 

Symmetric matrix

$$\Sigma = V^T D V$$

Diagonalization (SVD)

$$D = \begin{bmatrix} \lambda_1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \lambda_n \end{bmatrix} = \begin{bmatrix} \sigma_1^2 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \sigma_n^2 \end{bmatrix}$$

Eigenvalues  
Variances

$$\lambda_1 > \lambda_2 > \cdots > \lambda_n$$

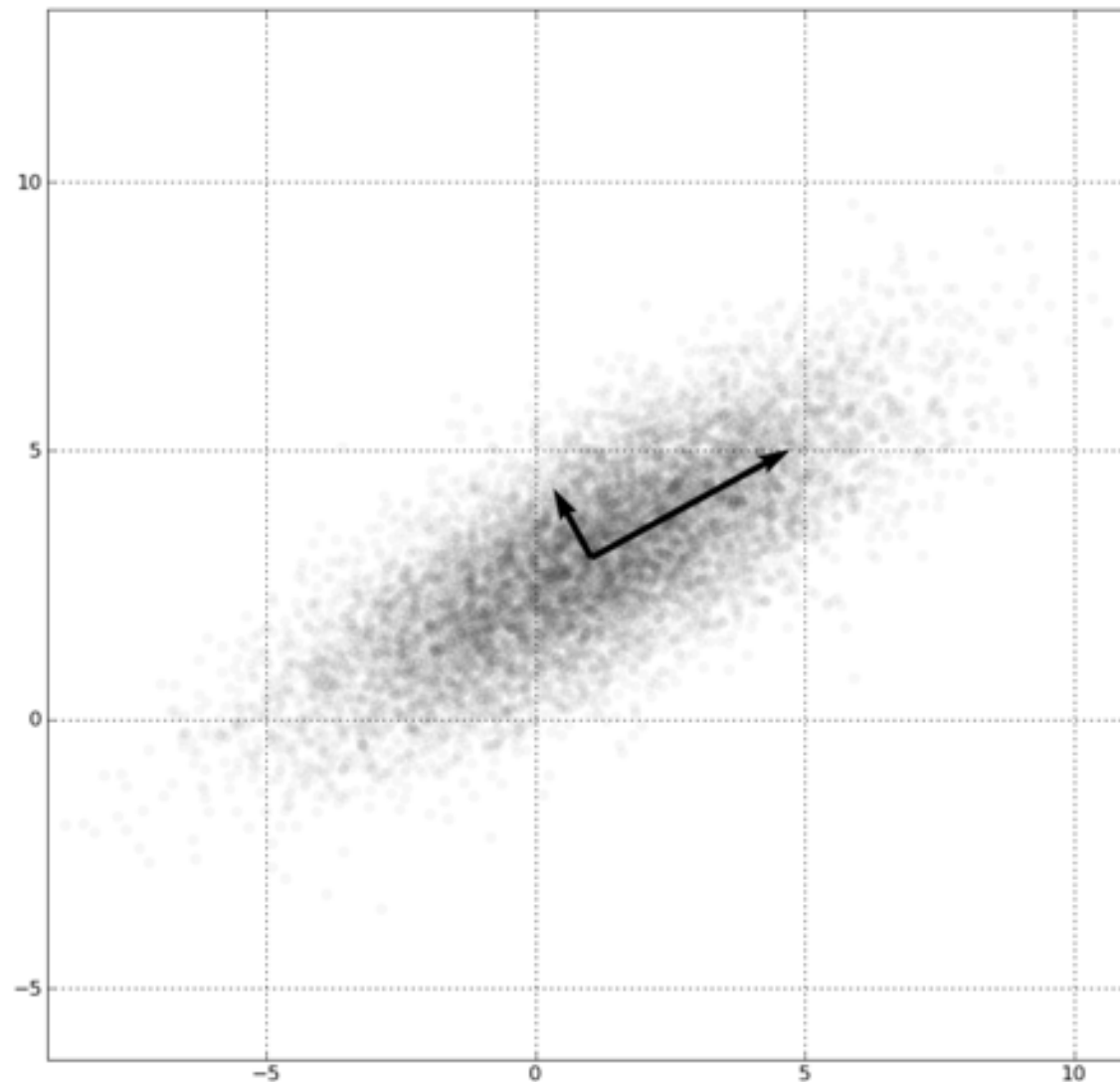
$$V = \begin{bmatrix} v_{1,1} & \cdots & v_{n,1} \\ \vdots & \ddots & \vdots \\ v_{1,n} & \cdots & v_{n,n} \end{bmatrix}$$

First principal component

Eigenvectors

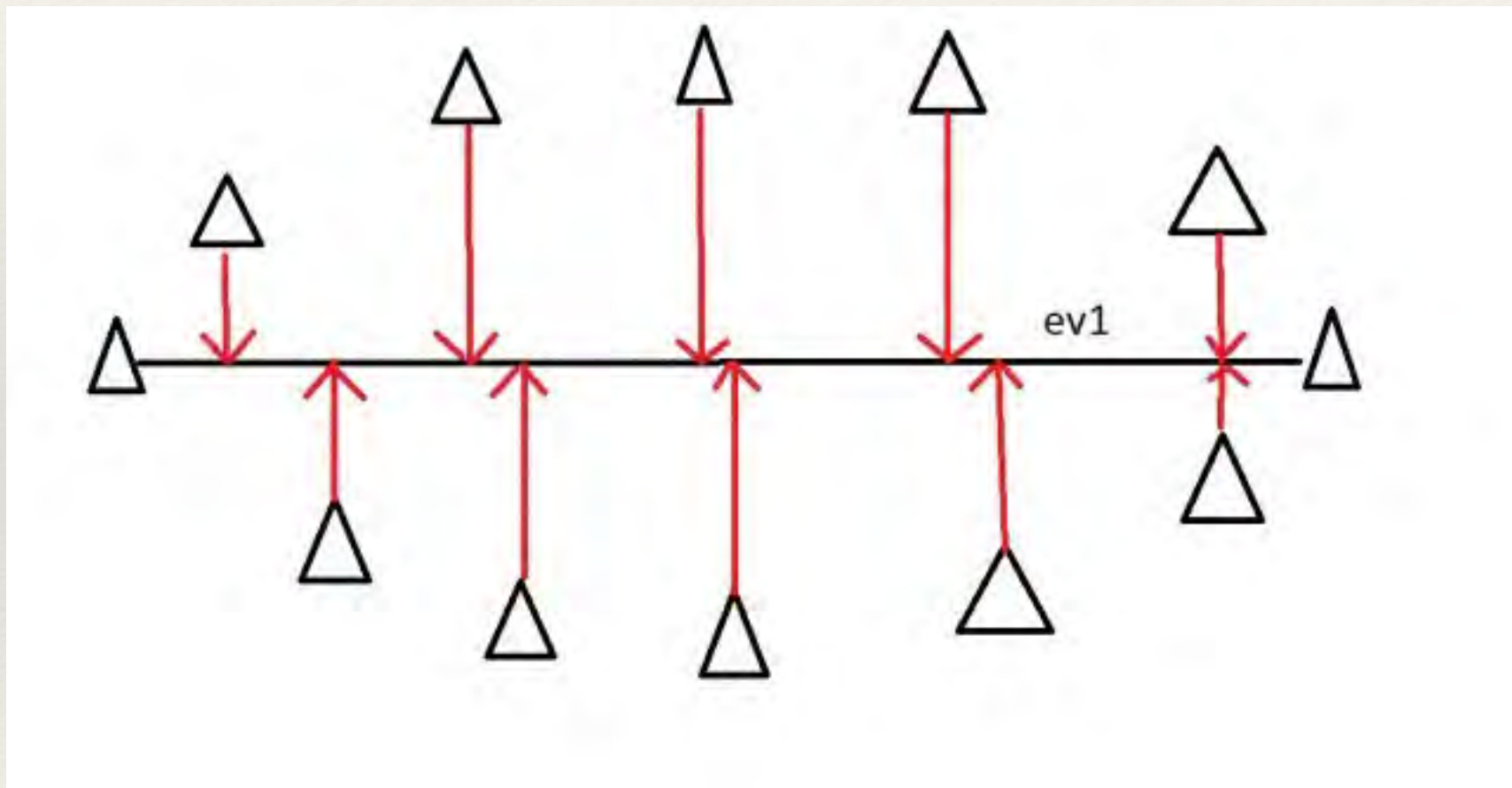


# Eigenvectors = Principal Components



First principal component = direction of greatest variance

# Least Squares





# Data Oriented Coordinates

