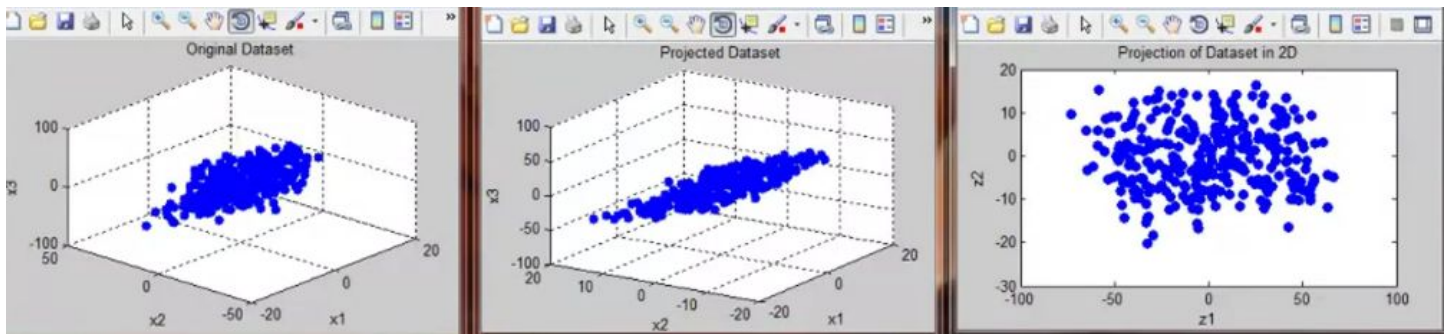


## **Plan**

- 1. Dimensional reduction - reduce the number of features to simplify data**
  - a. Improving performance**
  - b. Visualization**
- 2. Principal Component Analysis - PCA algorithm overview**
  - a. Understanding PCA problem**
  - b. PCA Algorithm**
  - c. Comparing PCA with linear regression**
- 3. Applying PCA - recommended ways to use PCA**
  - a. Reconstruction from compressed representation**
  - b. Choosing the number of principal components**
  - c. Advice for applying pca**



Reduce from  $n$ -dimension to  $k$ -dimension: Find  $k$  vectors  $u^{(1)}, u^{(2)}, \dots, u^{(k)}$  onto which to project the data, so as to minimize the projection error.

$$\text{Sigma} = \frac{1}{m} \sum_{i=1}^m (x^{(i)})(x^{(i)})^T$$

`[U,S,V] = svd(Sigma);`

`Ureduce = U(:,1:k);`

`z = Ureduce' * x;`

Typically, choose  $k$  to be smallest value so that

$$\frac{\frac{1}{m} \sum_{i=1}^m \|x^{(i)} - x_{approx}^{(i)}\|^2}{\frac{1}{m} \sum_{i=1}^m \|x^{(i)}\|^2} \leq 0.01 \quad (1\%)$$

“99% of variance is retained”

## Application of PCA

- Compression
  - Reduce memory/disk needed to store data
  - Speed up learning algorithm
- Visualization

### Questions:

1. What is dimensional reduction?
2. What's the main reasons to use PCA?
3. Which parameter illustrates percentage of retained variance?