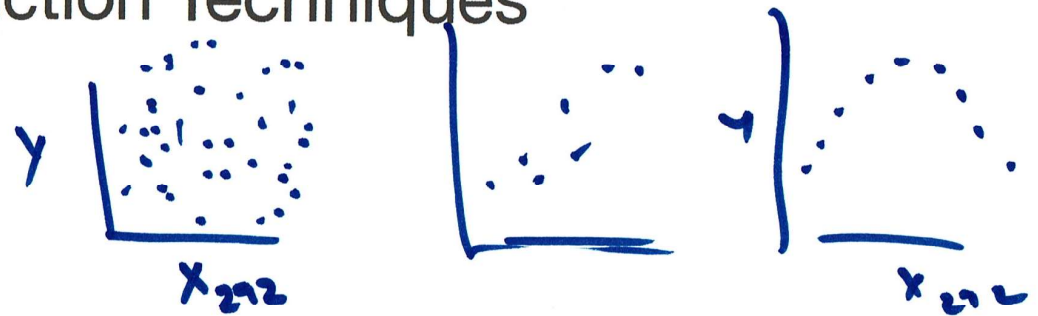


$$y = a + b_1 \cdot \square + b_2 \cdot \square \dots \dots \dots + b_{2000} \cdot \square$$

Dim. Reduction Techniques

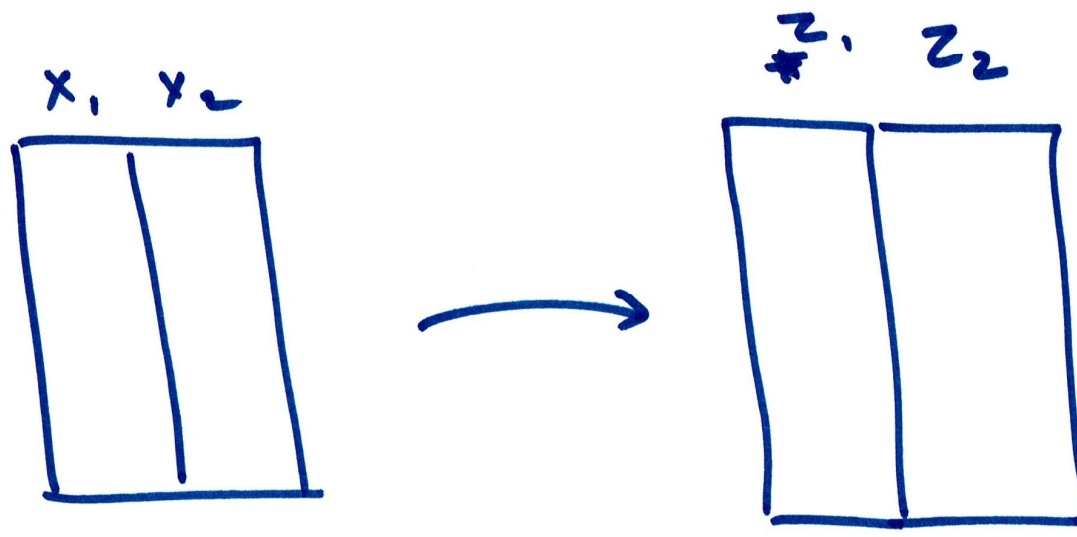


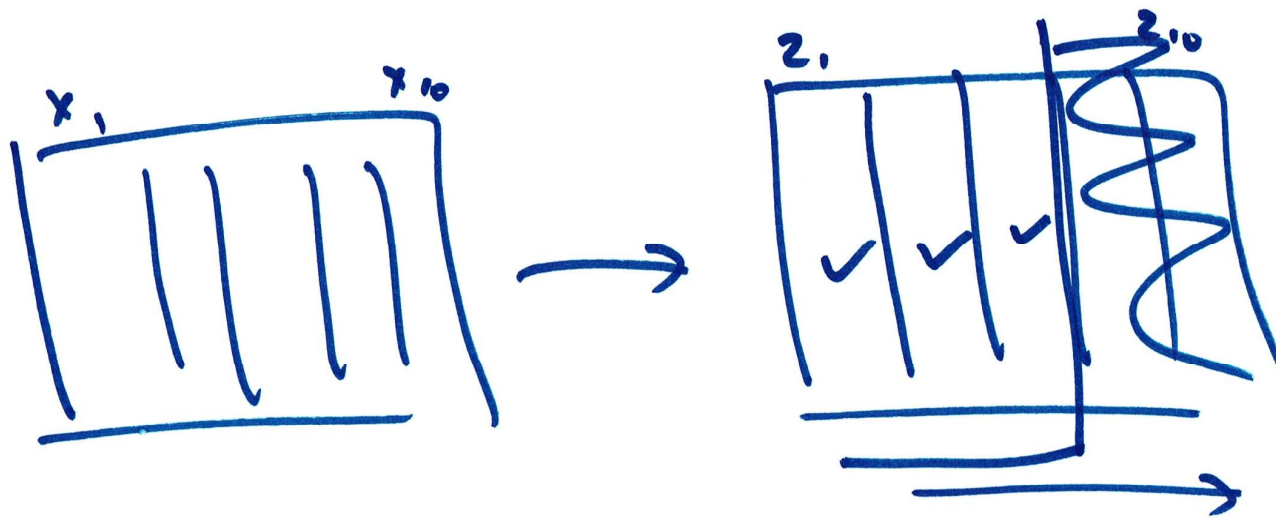
- Feature elimination

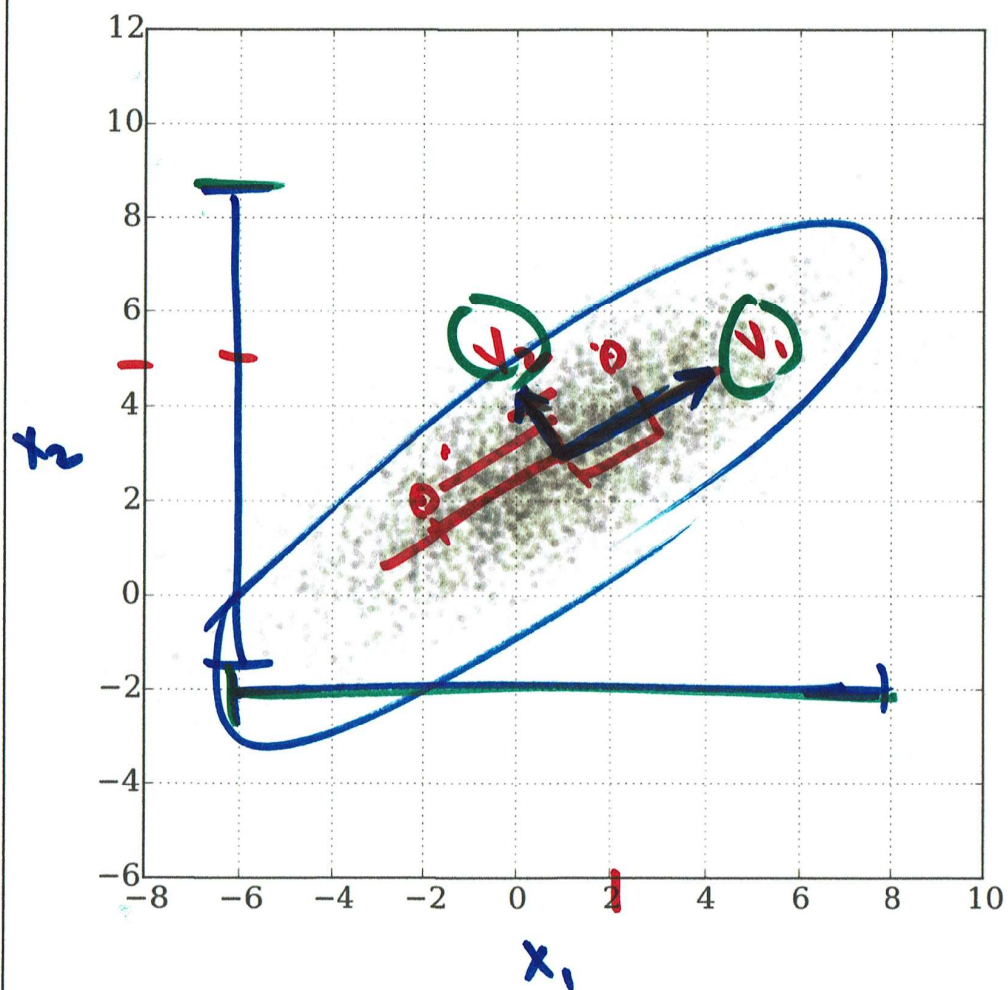
- Simply identify and remove variables (columns) that are not important
- The disadvantage is that we would gain no insight from those dropped variables and lose any information they contain

- Feature extraction

- Create a few new variables from the old variables
- **PCA** Principal Component Analysis: is the most popular feature extraction technique (linear)
- t-SNE (non-linear) ←







$$\begin{array}{c|c} x_1 & x_2 \\ \hline 2 & 5 \\ -2 & 2 \end{array} \rightarrow \begin{array}{c|c} 1 & 0.8 \\ -2 & 0.5 \end{array}$$

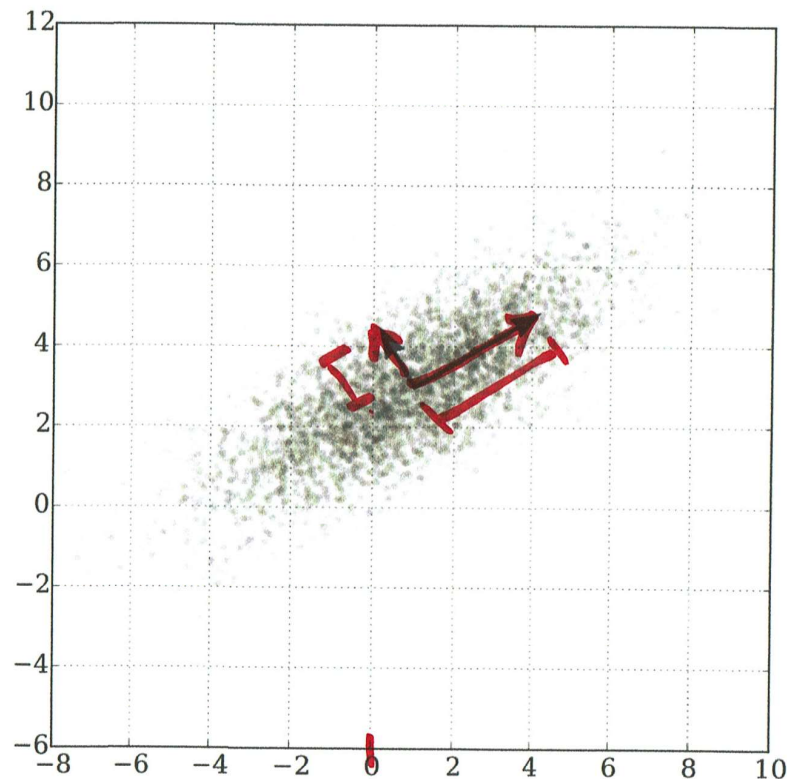
1.2 0.9 1.9 0.1

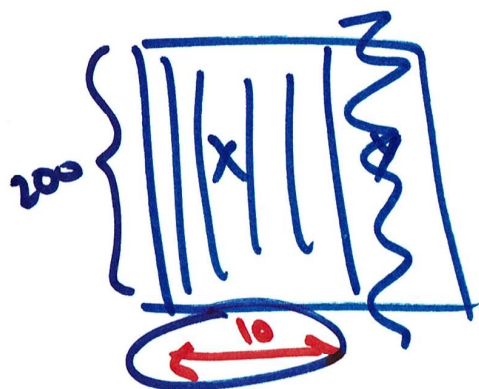
A diagram showing a horizontal band of red data points. A red line passes through the center of the band. A green line is also shown, slightly below the red line. Arrows point from the equations below to the red and green lines.

$$\begin{cases} Z_1 = \boxed{1.2} x_1 + \boxed{0.9} x_2 \\ Z_2 = \boxed{1.9} x_1 + \boxed{0.1} x_2 \end{cases}$$

PCA

- creates new variables using linear combinations of old variables
- is designed to create variables that are independent of one another
- also manages to tell us how important each of these new variables are
- this “importance”, helps us to choose how many variables we will use





$$X_{\text{new}} = \frac{X - \mu}{\sigma}$$

$$C_{\text{cov}} = \text{Cov}(X_{\text{new}})$$

Eigen decomposition

$$\begin{array}{c} \rightarrow \begin{array}{|c|c|c|c|} \hline e_1 & e_2 & \dots & e_{10} \\ \hline v_1 & v_2 & \dots & v_{10} \\ \hline \end{array} \end{array}$$

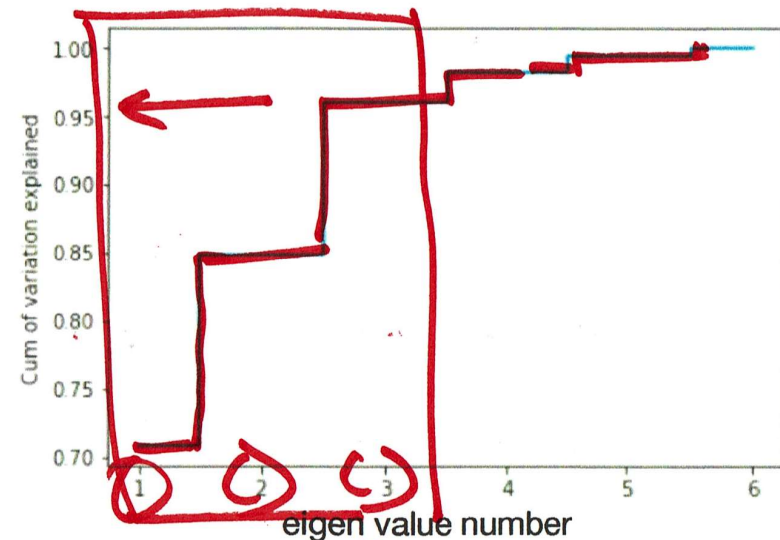
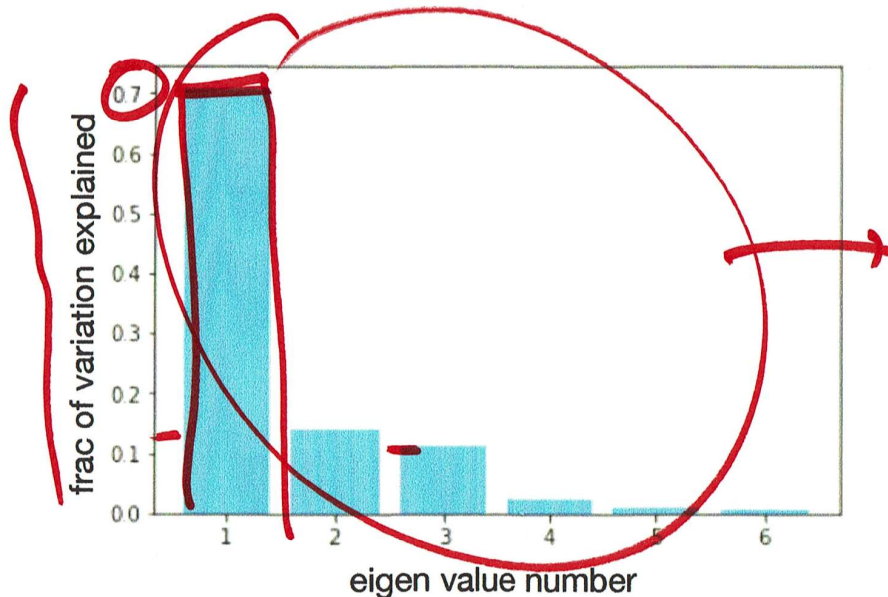
$$Z$$

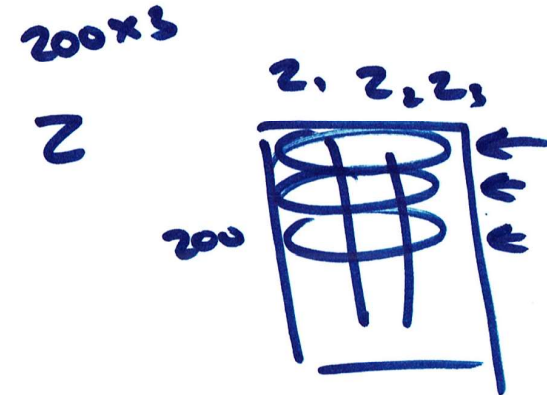
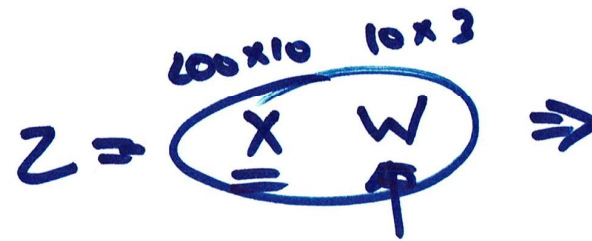
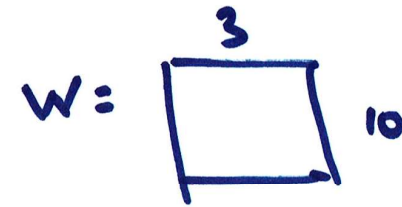
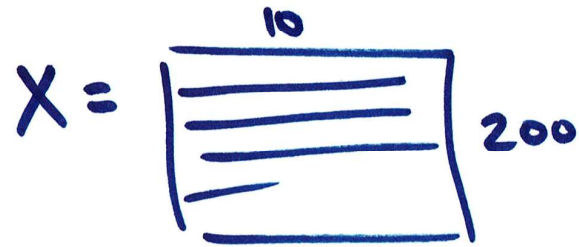
$$10 \times 3$$

$$W = \begin{pmatrix} v_1 & v_2 & v_3 \end{pmatrix}$$

- Scale the data and compute the covariance matrix
- Break the covariance matrix into magnitude and direction. Eigen Vectors and the Eigen Values of the covariance matrix can be thought of as the natural axis/directions and magnitudes along those axis, of the data
- The eigen values also can be used to calculate the percentage of variation explained by each component
- Sort in the eigen values in descending order and calculate the cumulative percentage of variation explained
- Pick the number of principal components you will use
- Transform to new variables

$$\frac{e_1}{\sum e_i}, \frac{e_2}{\sum e_i}, \frac{e_3}{\sum e_i}$$





$$S = 100 + 20 \underline{z_1} + 15 z_2 - 10 z_3$$

$$z_1 = \square x_1 + \square x_2 + \square x_3$$

$$z_2 = \dots$$

$$z_3 = \dots$$

1... x₁₀