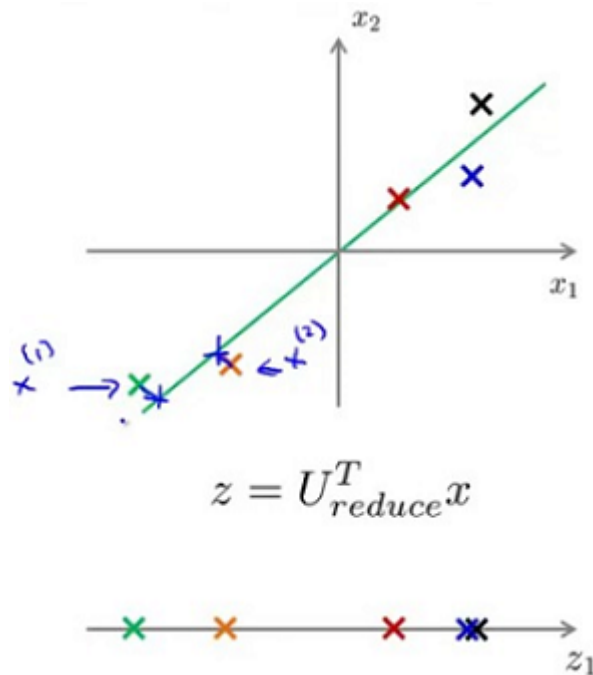


## Reconstruction from Compressed Representation

Previously we were talking about PCA as a compression algorithm where we may have say, 1000-dimensional data and compress it to 100-dimensional feature vector. So, if this is a compression algorithm, there should be a way to go back from this compressed representation back to an approximation of our original high-dimensional data.

## Reconstruction of our high-dimensional data

Say we have an example as follows



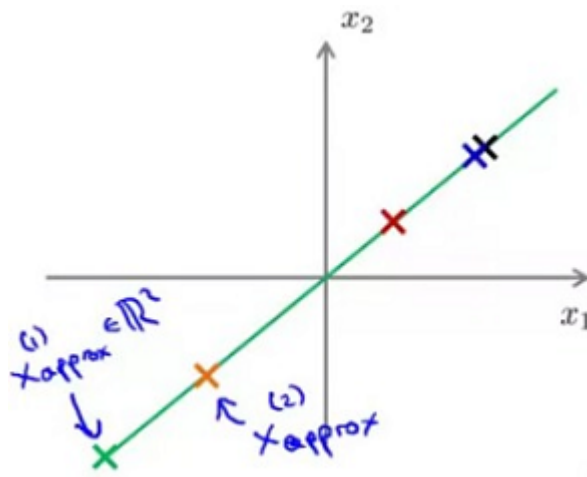
- We have our examples  $(x^{(1)}, x^{(2)}, \dots, x^{(n)})$ 
  - Project onto  $z$ -surface
- Given a point  $z^{(1)}$ , how can we go back to the 2D space  $z \in \mathbb{R} \rightarrow x \in \mathbb{R}^2$ ?
  - Considering:  $z = (U_{\text{reduce}})^T * x$
- To go in the opposite direction we must do:
  - $X_{\text{approx}} = U_{\text{reduce}} * z$
  - To consider dimensions:
    - $U_{\text{reduce}} = [n \times k]$
    - $z = [k \times 1]$
    - So  $X_{\text{approx}} = [n \times k] * [k \times 1] = [n \times 1]$

$$z \in \mathbb{R} \rightarrow x \in \mathbb{R}^2$$

$$X_{\text{approx}} \stackrel{\approx x}{=} \underbrace{U_{\text{reduce}}}_{n \times k} \cdot \underbrace{z}_{k \times 1}$$

$$\underbrace{\quad}_{n \times 1}$$

So this creates the following representation:



And that's a pretty decent approximation to the original data. So that's how we go back from your low dimensional representation  $z$ , back to an uncompressed representation of the data, and we also call this process reconstruction of the original data where we think of trying to reconstruct the original value of  $x$  from the compressed representation.

We lose some of the information (i.e. everything is now perfectly on that line) but it is now projected into 2D space

**Video Question:** Suppose we run PCA with  $k = n$ , so that the dimension of the data is not reduced at all. (This is not useful in practice but is a good thought exercise). Recall that the percent / fraction of variance retained is

given by:  $\frac{\sum_{i=1}^k S_{ii}}{\sum_{i=1}^n S_{ii}}$ . Which of the following will be true? Check all that apply.

$U_{\text{reduce}}$  will be an  $n \times n$  matrix.

$x_{\text{approx}} = x$  for every example  $x$ .

The percentage of variance retained will be 100%.

- We have that  $\frac{\sum_{i=1}^k S_{ii}}{\sum_{i=1}^n S_{ii}} > 1$