

Steps for Principal Component Analysis

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- Standardisation
- Finding the Covariance Matrix
- Finding the directions of maximum variance -
Eigenvectors and Eigenvalues
- Selecting the Principal Components

Standardisation

Why is standardisation required?

- Different features have different ranges and scales

Consider the features in the famous breast cancer dataset

	Min	Max
radius (mean):	6.981	28.11
texture (mean):	9.71	39.28
perimeter (mean):	43.79	188.5
area (mean):	143.5	2501.0
smoothness (mean):	0.053	0.163
compactness (mean):	0.019	0.345
concavity (mean):	0.0	0.427
concave points (mean):	0.0	0.201
symmetry (mean):	0.106	0.304



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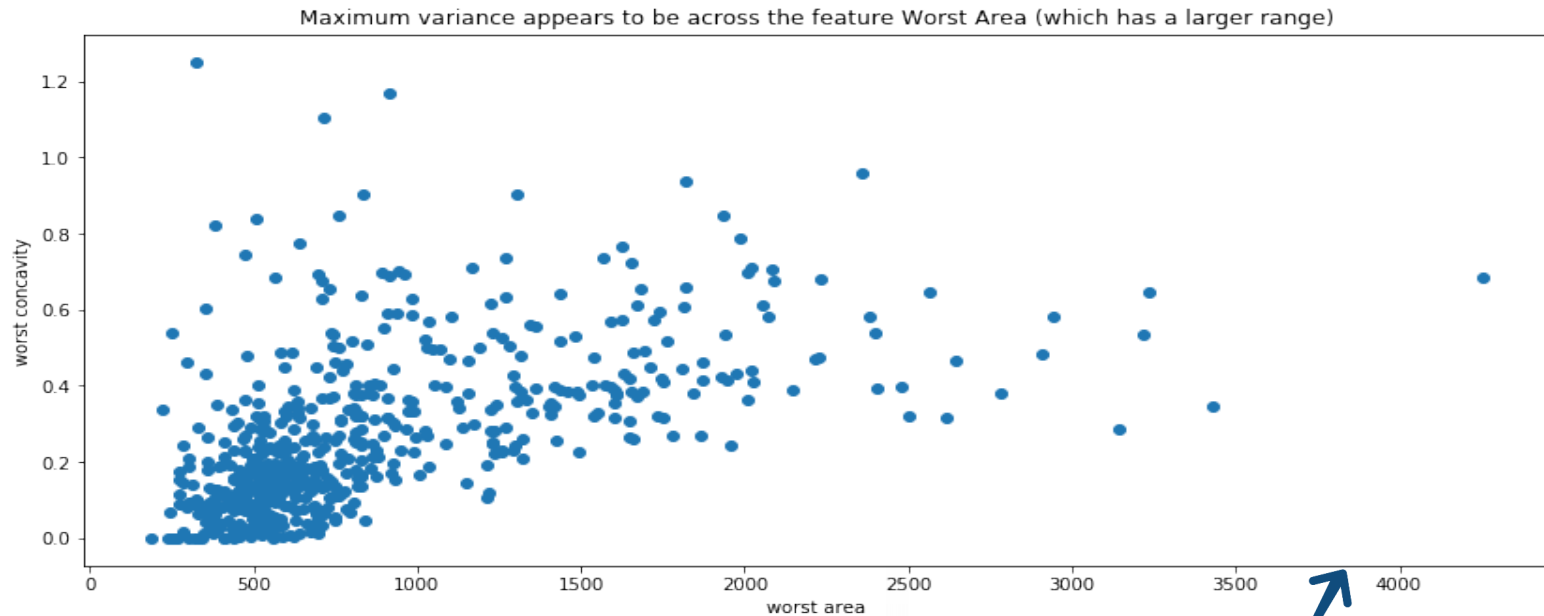
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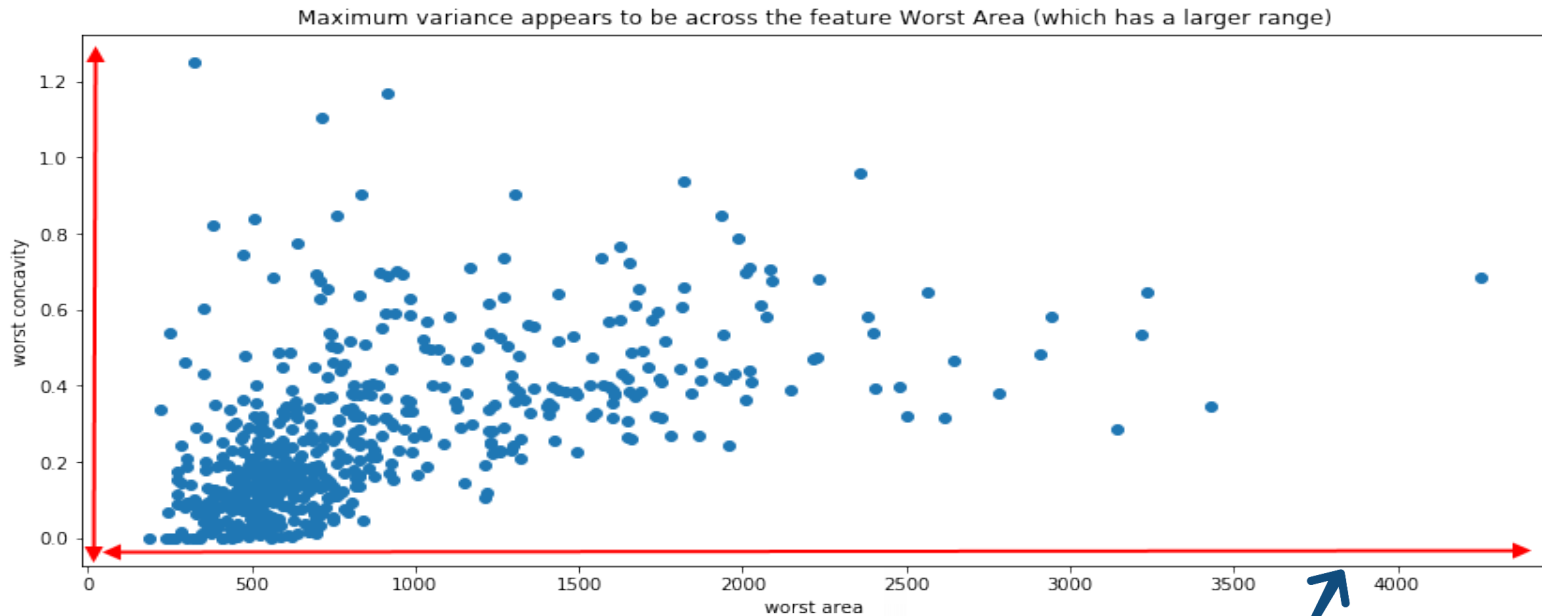
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- A feature with a larger range will have a higher variance.



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Standardisation is used so that each feature contributes equally to the PCA algorithm.

How to standardise a value?

Mathematically, the standardised value of a value x is given by:

$$x_{new} = \frac{x - \text{mean}(x)}{\text{std dev}(x)}$$

Properties of Standardised Data

Resultant features obtained after standardisation have the following properties:

- Distributed with mean = 0
- Distributed with variance = 1

The data is said to be **column-standardised**.

Thank
You!