# Theoretical and practical metagenomic approaches to viral discovery

Practical Session: Dimension Reduction and Clustering

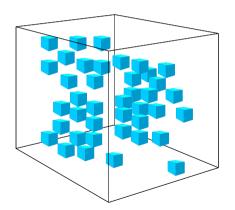
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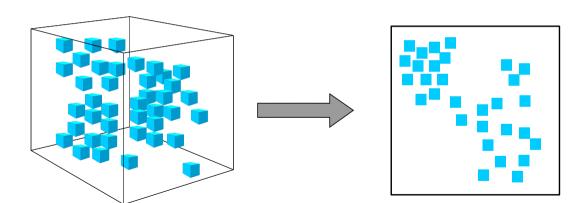
# **Dimension Reduction**





PCA Classification

# **Dimension Reduction**





## PCA with scikit-learn



#### A QUICK EXAMPLE



#### NORMALIZE YOUR DATA

x = StandardScaler().fit\_transform(x)

```
features = ['sepal length (cm)', 'sepal width (cm)', 'petal length (cm)', 'petal width (c

# Here we seperate features from the target column

x = df.loc[:, features].values

y = df.loc[:,['target']].values

# Standardizing the only the features
```



#### APPLYING PCA



### Virus Classification with PCA



#### OTHER APPLICATION OF PCAS

#### Dimension Reduction in Machine Learning

Normally, you would like to use PCA (or any other method for dimension reduction) to **speed up** your machine learning algorithm. Instead of learning many instances with over 700 features (e.g. our handwritten letter example from earlier), we can **reduce the number of features**, by only taking the most **important combinations** of features into account.



#### EXERCISE / HOMEWORK

Have a look at the viral data fasta.

There are some known viruses, but also some unknown viruses inside. Try to use k-mer frequencies as features and cluster the sequences based on that.

Can you roughly classify the unknown viruses?

You do not have to do this in Python. If you are more familiar with, for example, R, feel free to use this as well.

I am just a little bit Python addicted. :)



PCA

## COFFEE BREAK



