



# Library Academy

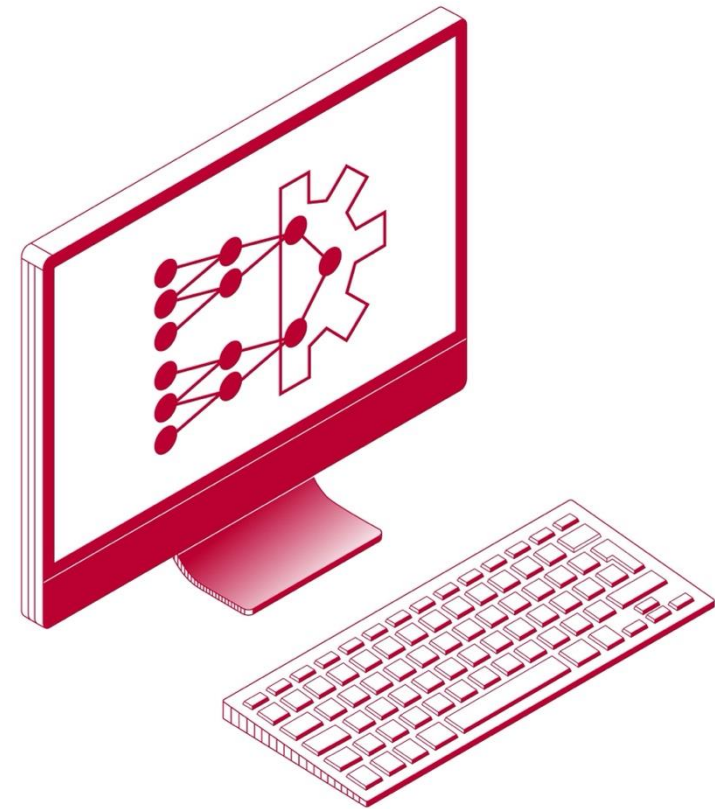
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21-01-2026



# Ontdek wat data vertellen: Aan de slag met Machine Learning in JASP

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Library Academy 2026



# Outline

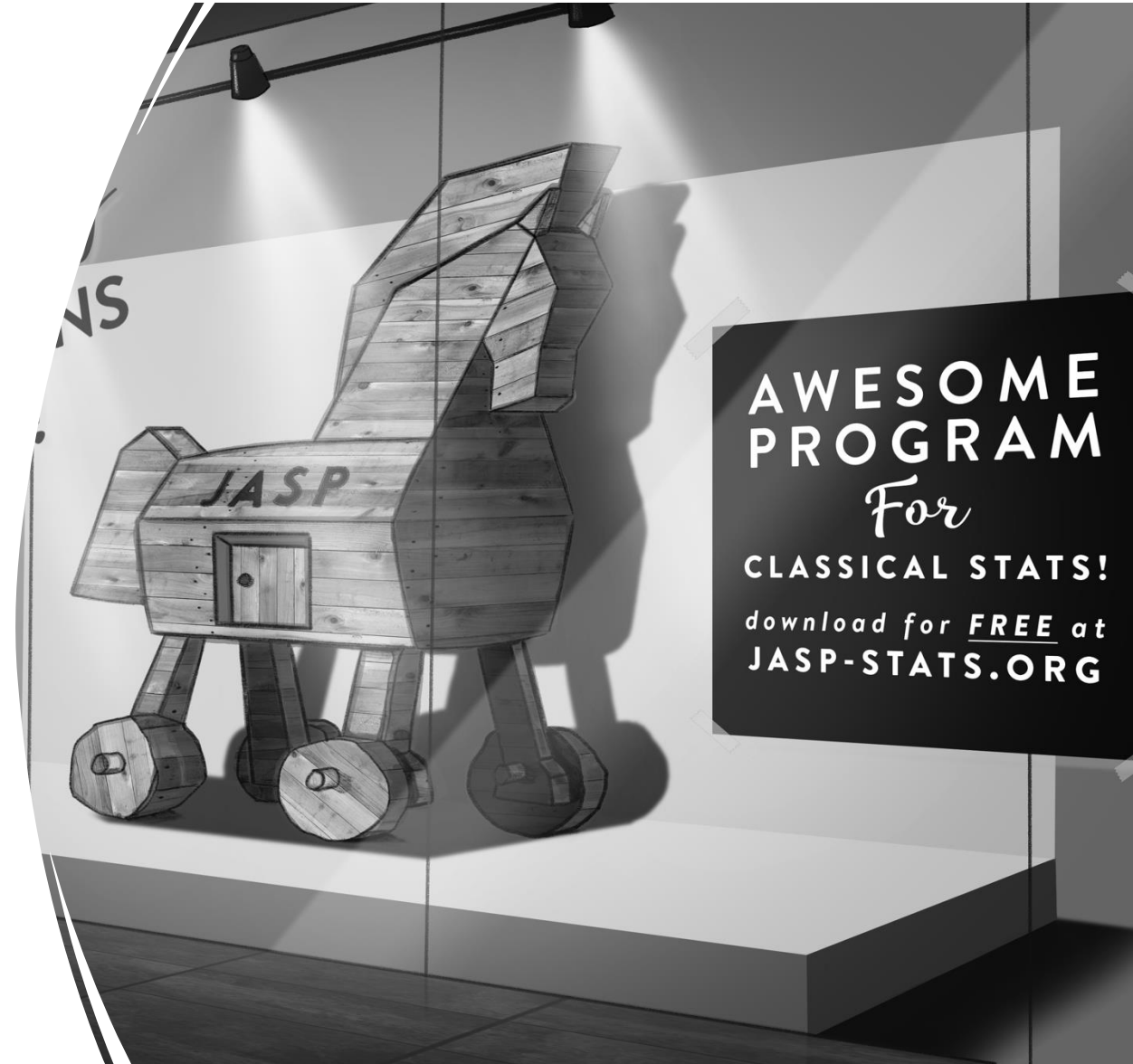
- Basics
  - What is JASP?
  - Regression
- Machine Learning
  - General philosophy
  - K-nearest neighbors
- Exercises



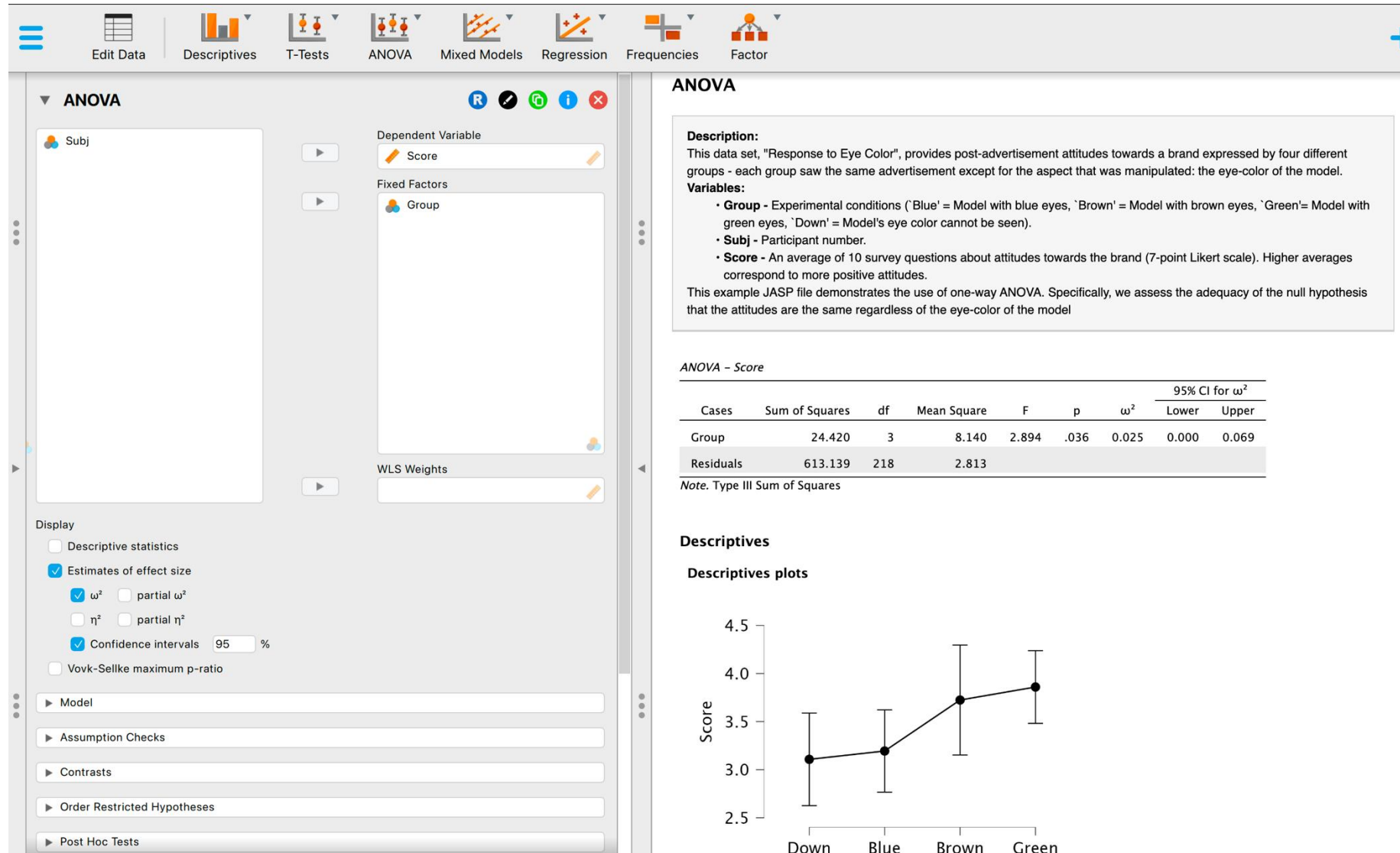
[edu.nl/knhfd](https://edu.nl/knhfd)

# What is JASP?

- Developed at UvA over the past 10 years, funded by research grants (NWO/EU)
- Graphical user interface for conducting frequentist and Bayesian statistics
- <https://jasp-stats.org/>
- Open-source → forever free!



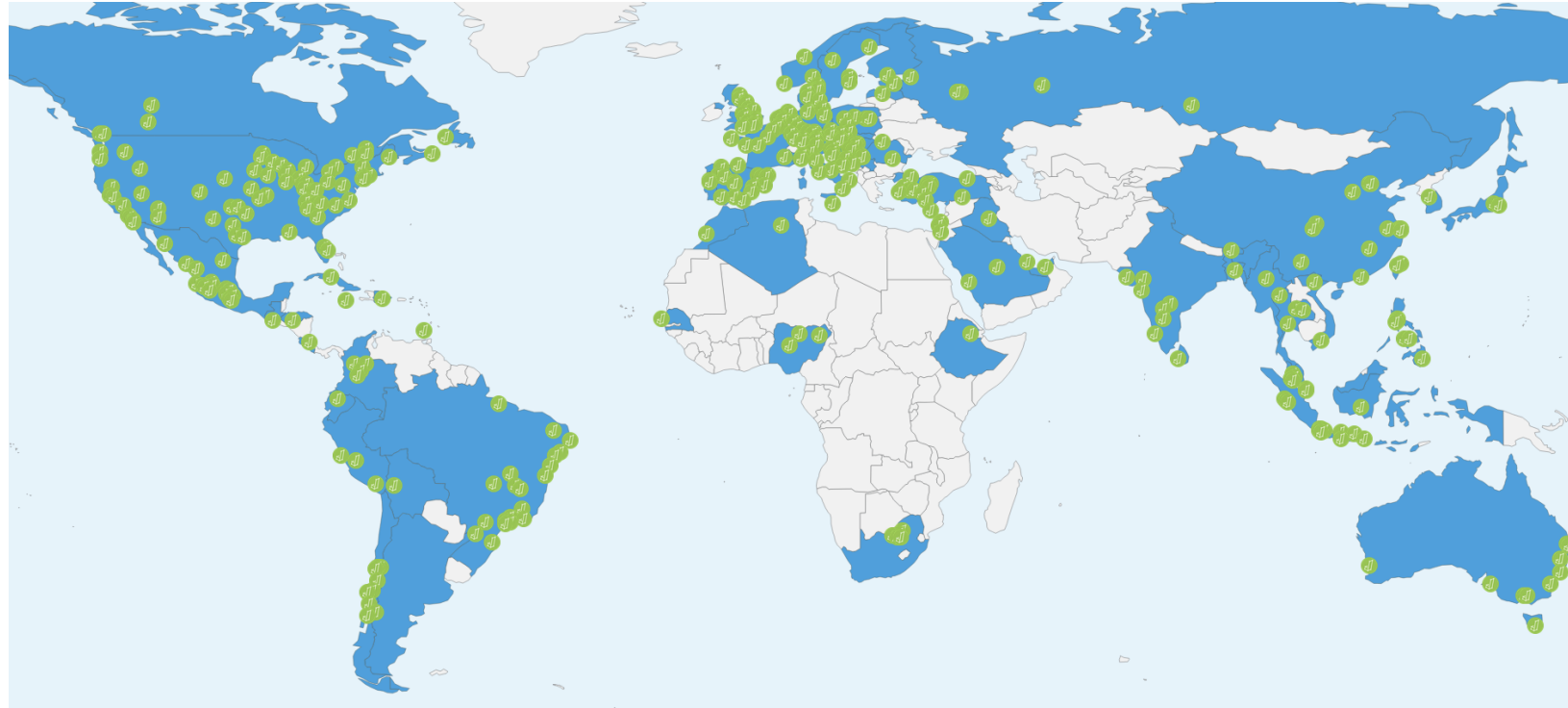
# What is JASP?





# What is JASP?

- Used at 374 universities across 76 countries
- 100,000 monthly downloads



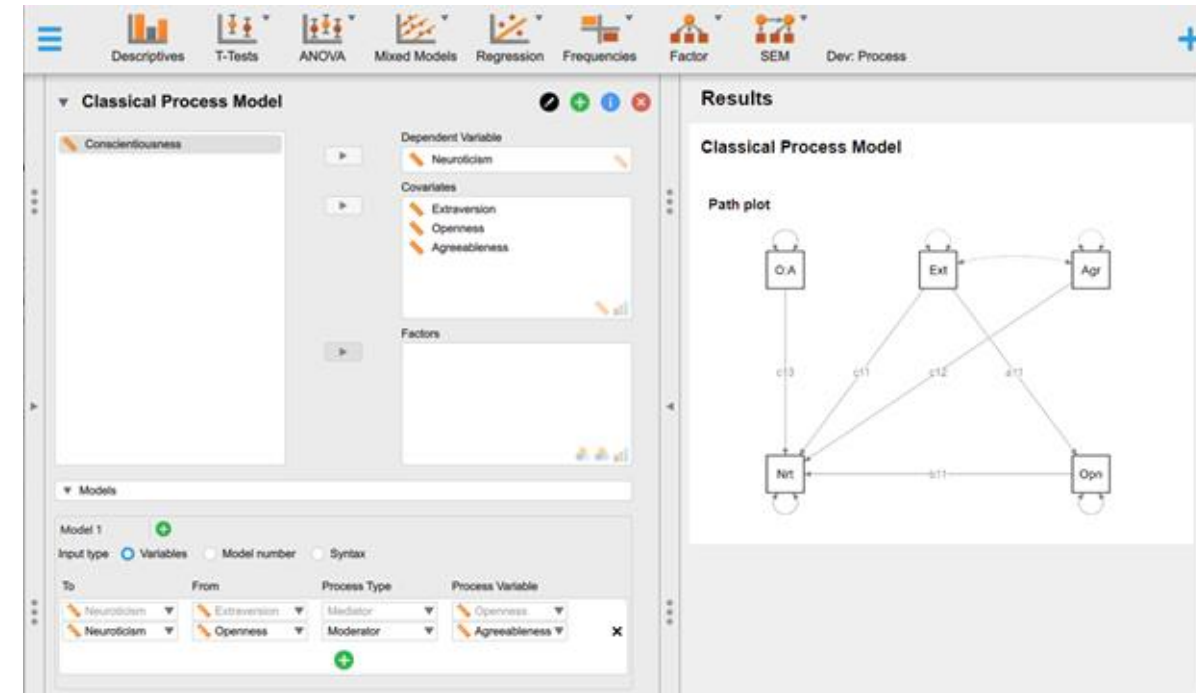
<https://jasp-stats.org/teaching-with-jasp/#worldmap>

# Features

- [Website overview](#)
- [JASP vs. SPSS feature comparison](#)
- Data formats: .sav, .xls, .txt, .csv, .ods, .tsv, .dta, .por, .sas7bdat, .sas7bcat, and the .jasp format
- APA tables
- OSF integration
- R console
- Compute columns
- Filtering

# Feature Roadmap

- Full syntax mode ([blog about the first implementation](#))
- Better data manipulation
- Select filters





# Other Handy Resources












- [How to Use JASP – Inventory of blogs/videos/gifs for frequentist and Bayesian analyses](#)
- [JASP YouTube page](#)
- [The JASP Video Library](#)
- [Step By Step Guide: 1. Bayesian One-Way ANOVA](#) and the [full playlist](#)
- JASP on Bluesky - <https://bsky.app/profile/jaspstats.bsky.social>
- JASP forum - <https://forum.cogsci.nl/index.php?p=/categories/jasp-bayesfactor>
- Found a bug? Please report on Github: <https://github.com/jasp-stats/jasp-issues/issues>
- [JASP Verification Project](#)
- More JASP workshops: <https://jasp-stats.org/workshop/>

# JASP Literature

- [The JASP Data Library](#)
- [Discovering Statistics Using JASP](#)
- [Learning Statistics with JASP: A Tutorial for Psychology Students and Other Beginners by Danielle J. Navarro, David R. Foxcroft, and Thomas J. Faulkenberry](#)
- [Statistics of Doom by Erin Buchanan](#)
- [Statistical Analysis in JASP. A Guide for Students by Mark Goss-Sampson](#)
- [Quantitative Analysis with JASP open-source software by Chris Halter](#) (amazon)




# Data Management

		 Analyses	 Synchronisation	 Resize Data	 Insert	 Remove
		 Name	 Instrument	 Current member	 Headbanging intensity	
1		Lars Ulrich	Drums	Yes 1	Light	1
2		James Hetfield	Guitar	Yes 1	Heavy	3
3		Kirk Hammett	Guitar	Yes 1	Light	1
4		Rob Trujillo	Bass	Yes 1	Moderate	2
5		Jason Newsted	Bass	No 0	Heavy	3






# The Variable View

Name:  Long name:




Column type:  Nominal ▼ Description:

Computed type: Not computed ▼

Label editor Missing values

	Filter	Value	Label
	✓	Lars Ulrich	Lars Ulrich
	✓	James Hetfield	James Hetfield
	✓	Kirk Hammett	Kirk Hammett
	✓	Rob Trujillo	Rob Trujillo
	✓	Jason Newsted	Jason Newsted


# Variable Types

- Scale 
  - Numbers (e.g., 7, 0, 120, 8.5)
- Nominal 
  - Categories (e.g., 'Control group', 'Experimental group')
- Ordinal 
  - Ordered values (e.g., 'Dislike', 'Neutral', 'Like')



# Variable Settings

Name:

Column type:  Ordinal ▼

Computed type:  ▼

Label editor Missing values

1  
N ↓

1  
N ↺

↑  
↓

▲

▼

Filter	Value	Label
✓	1	Light
✓	2	Moderate
✓	3	Heavy

# Computing a New Variable

Name:  Long name:

Column type:  Description:

Computed type:

Computed column definition Label editor Missing values

$+ - * \div / ^ \sqrt \% = \neq < \leq > \geq \wedge \vee | \neg$

☐ Name ☐ Instrument ☐ Current member ☐ Headbangi... ntensity

☐ Net worth... million)  
☐ Songs

Computed columns code applied

☐ Converting types

$|y|$   
 $\sigma_y$   
 $\sigma^2_y$   
 $\Sigma y$

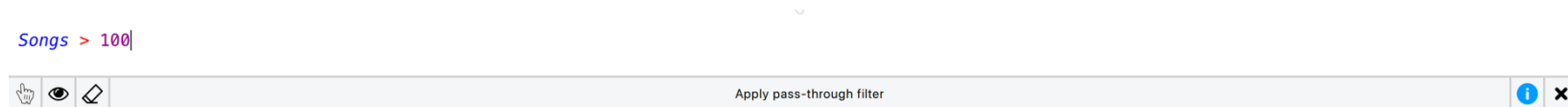


# Filtering Data

- Using Variable Settings
- Using the Filter functionality
  - Drag and drop





- R-mode





# Descriptives – input window


▼ **Descriptive Statistics** R ✎ 📄 i ✕

 Name

 Instrument

 Current member


 Net worth (\$ million)


 Net worth per song

↓  
A  
Z


▶

Variables

 Songs

 Headbanging intensity

Split
















☐ Transpose descriptives table

▶ Statistics

# Descriptives – output window

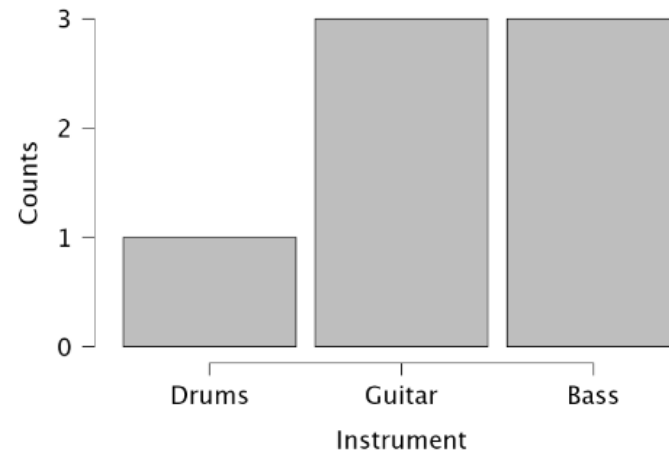
## Distribution Plots

B I U   $f_x$   $\langle / \rangle$    Normal      $x_2$   $x^2$     Normal    $I_x$

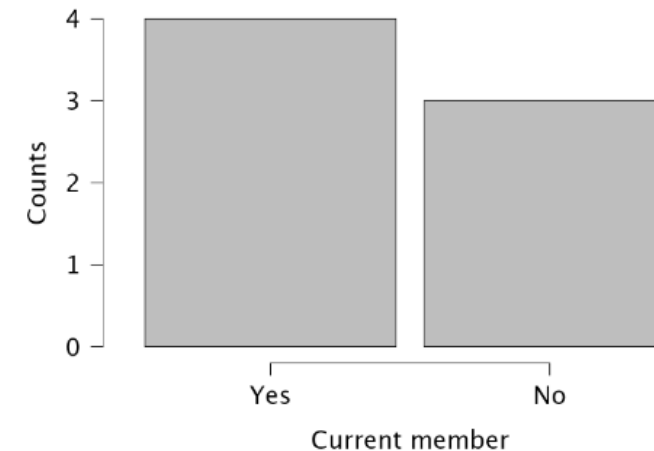
We can have fancy  $LaTeX$  formula's in here, [link cute cat video's](#), or insert a drumkit 

Below are two distribution plots outlining the members of Metallica. On the left, we see the various instruments being played and their frequencies, and on the right we see how many members are still active, and how many left the band.

Instrument



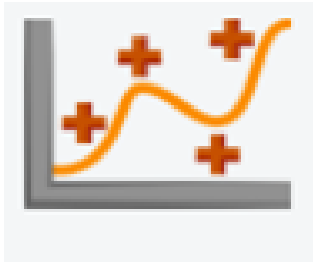
Current member







# There are lots of different machine learning algorithms



**Regression**

- Boosting
- Decision tree
- K-nearest neighbors
- Linear
- Neural network
- Random forest
- Regularized linear
- Support vector machine



**Classification**

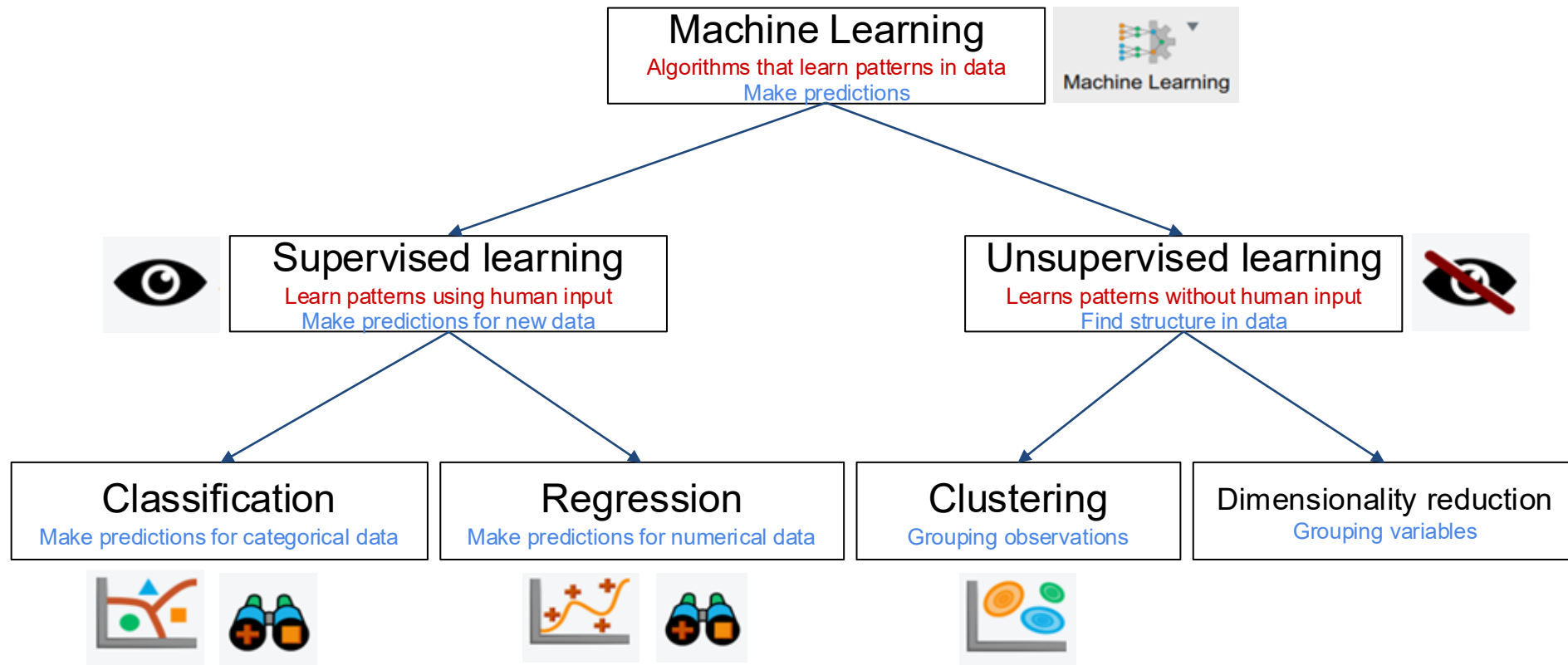
- Boosting
- Decision tree
- K-nearest neighbors
- Linear discriminant
- Logistic/Multinomial
- Naive Bayes
- Neural network
- Random forest
- Support vector machine



**Clustering**

- Density-based
- Fuzzy c-means
- Hierarchical
- Model-based
- Neighborhood-based
- Random forest

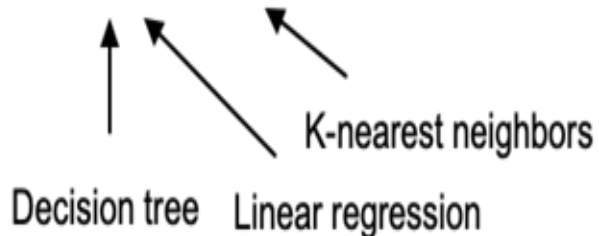
# Machine learning is about finding patterns, the goal is making predictions



- Some algorithms are 'black boxes'

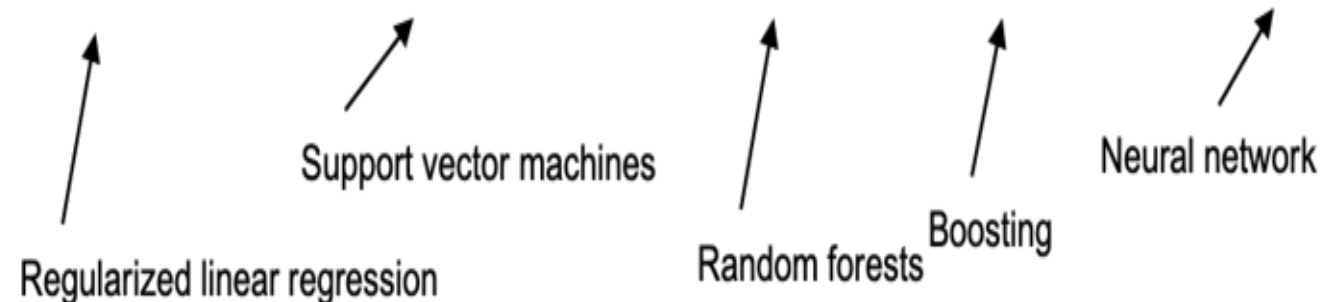


- Understanding all algorithms is not the most important, understanding how to evaluate their results is

**Explainable****Not explainable**

Decision tree   Linear regression   K-nearest neighbors

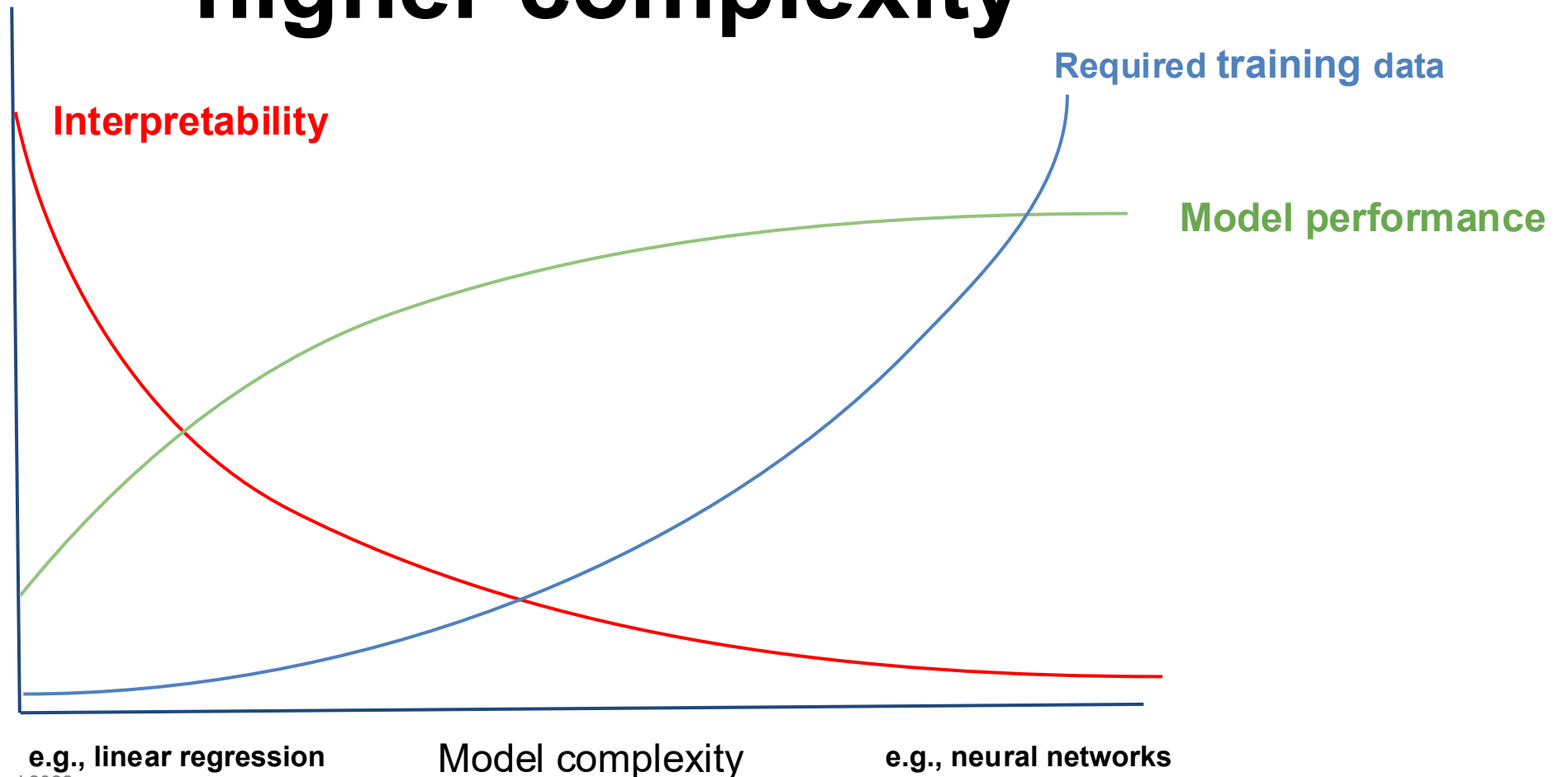
This section shows three algorithms (Decision tree, Linear regression, and K-nearest neighbors) with arrows pointing to the green portion of a horizontal gradient bar that transitions from green on the left to red on the right.



Regularized linear regression   Support vector machines   Random forests   Boosting   Neural network

This section shows five algorithms (Regularized linear regression, Support vector machines, Random forests, Boosting, and Neural network) with arrows pointing to the red portion of the same horizontal gradient bar.

# Better performance (often) comes with higher complexity



# Learning objectives for today

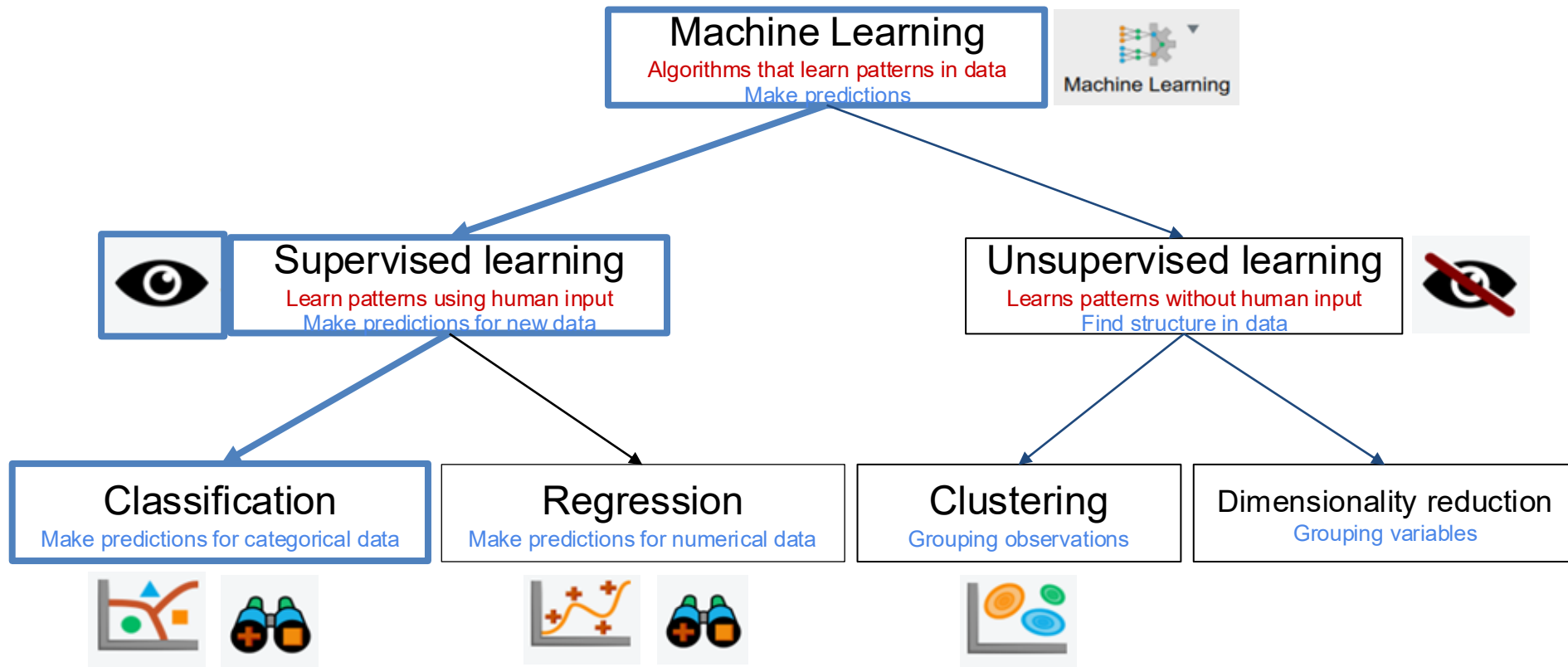
- Train machine learning algorithms (regression, classification, clustering) using JASP
- Understand how to evaluate the quality of these algorithms
- Apply these algorithms to predict new data

# Learning objectives for today

- Train machine learning algorithms (regression, **classification**, clustering) using JASP
- Understand how to evaluate the quality of these algorithms
- Apply these algorithms to predict new data



# Machine learning is about finding patterns, the goal is making predictions



# A typical classification workflow

1. Training the classification model
2. Optimizing the model (optional)
3. Evaluating the quality of the predictions of the model
4. Apply the model to new data

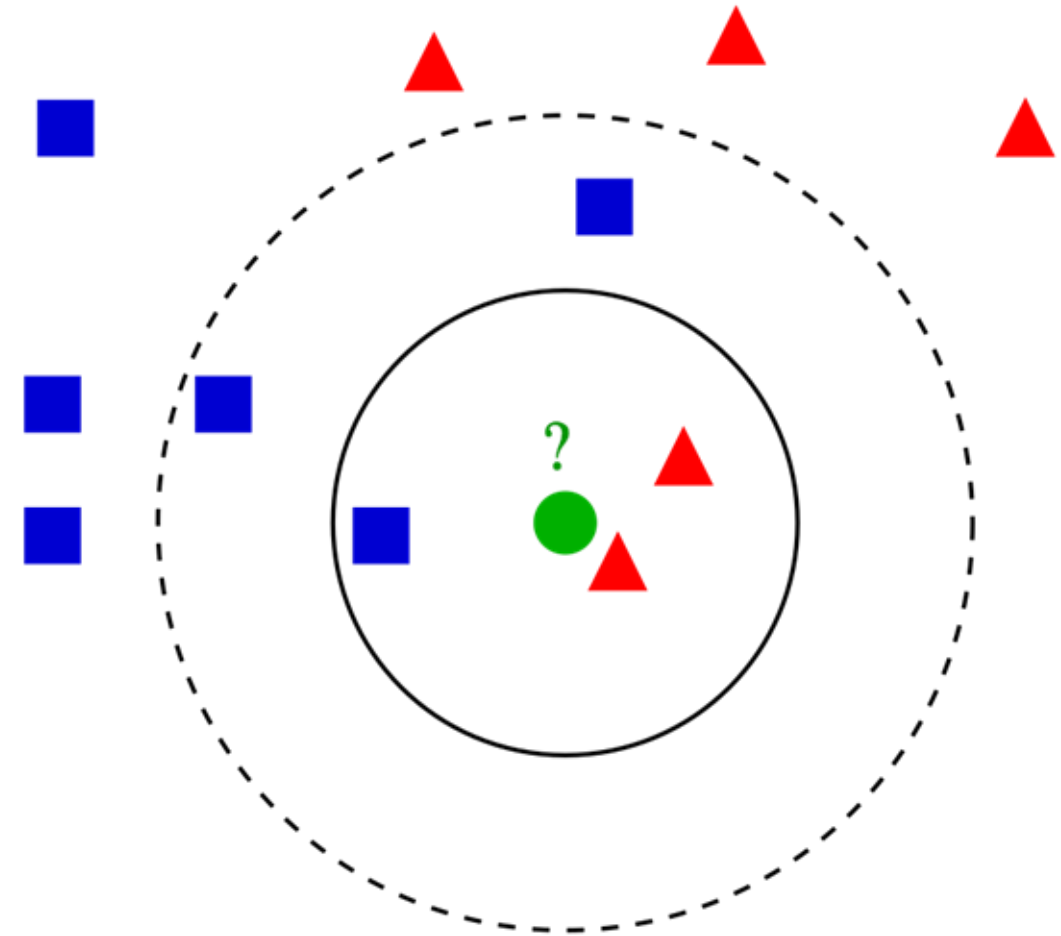
# You typically split your dataset in 2 or 3 parts

1. Training set (typically 80%): Used for training the model
2. Validation set: Used for optimizing the model (optional)
3. Test set (typically 20%): Used for evaluating the model

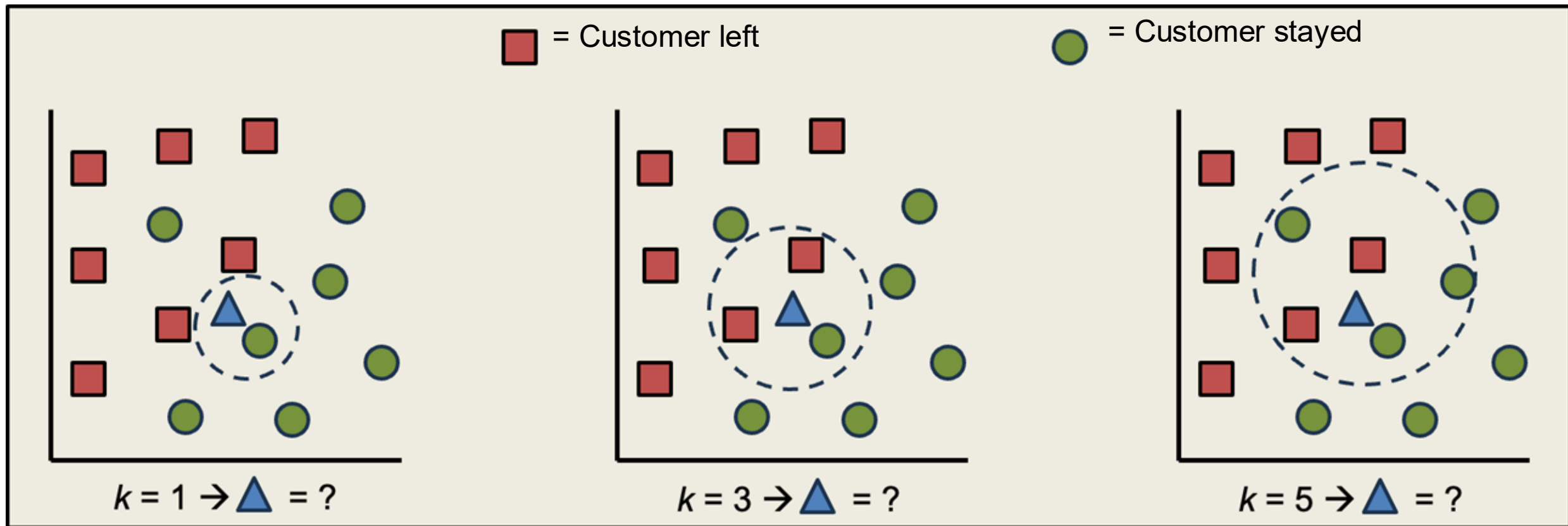


# Algorithm: K-nearest neighbors

- Finds the  $k$  most similar observations in the training set and uses this as a basis for prediction.
- For a new observation you find the  $k$  most similar observations and **take the most occurring category**.



# Algorithm: K-nearest neighbors



## Exercise: K-nearest neighbors

Using pen and paper, draw up the data as follows:

- Number of products sold on the horizontal axis,
- Number of years as a customer on the vertical axis,
- Use 😊 for a customer who stayed,
- Use 😞 for a customer who left

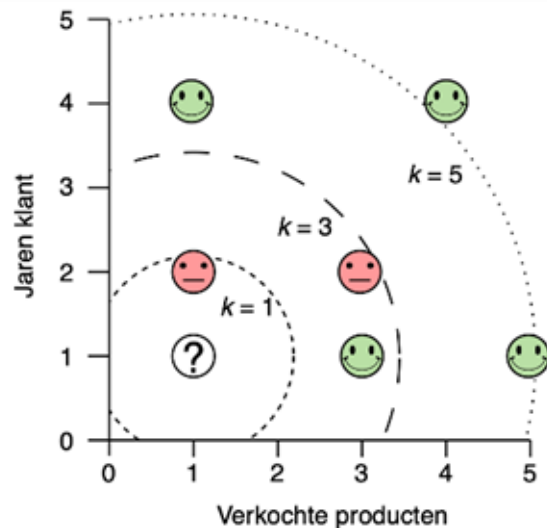
What is your prediction with  $k = 1, 3$  and  $5$  when products sold = years as customer = 1? 😊 😞 😞?

What is your prediction with  $k = 1, 3$  and  $5$  when products sold = 5 and years customer = 3 😊 😞 😞?

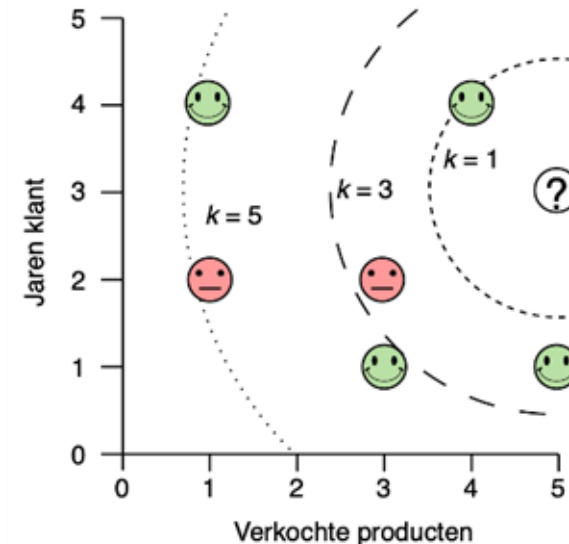
Sold products	Years customer	Outcome
1	4	Stayed
4	4	Stayed
1	2	Left
3	2	Left
3	1	Stayed
5	1	Stayed

## Exercise: K-nearest neighbors

- What is your prediction with  $k = 1, 3$  and  $5$  when products sold = years as customer = 1? 😊 😐 or 😞? 😊 or 😞?
- What is your prediction with  $k = 1, 3$  and  $5$  when products sold = 5 and years customer = 3 😊 or 😞?



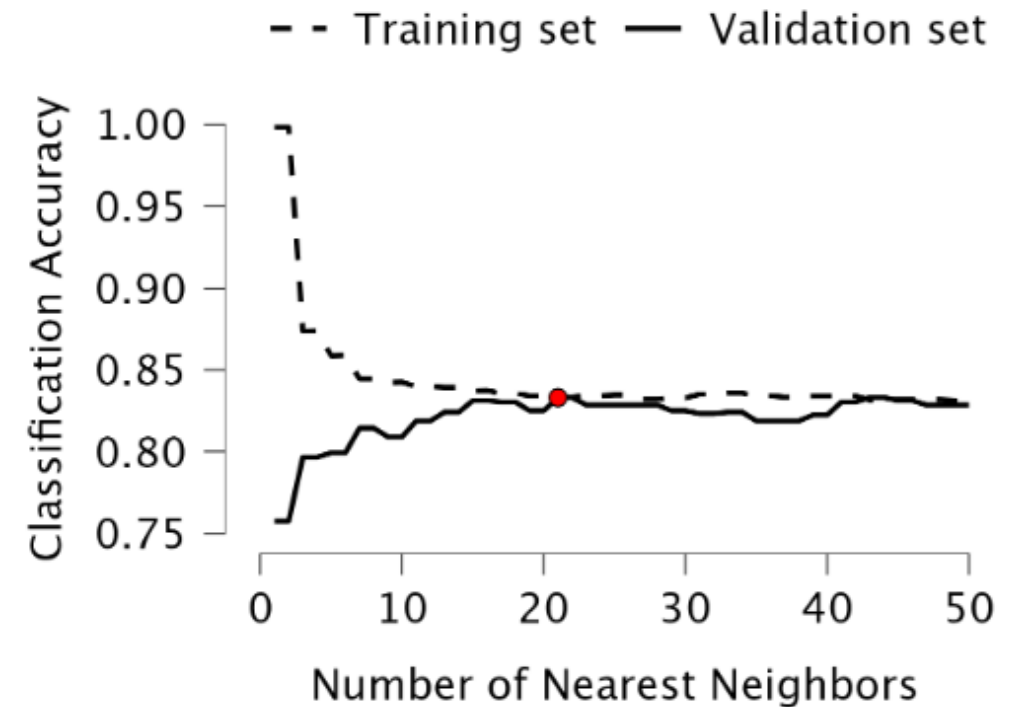
$k = 1 \rightarrow$  😞  
 $k = 3 \rightarrow$  😞  
 $k = 5 \rightarrow$  😊



$k = 1 \rightarrow$  😊  
 $k = 3 \rightarrow$  😊  
 $k = 5 \rightarrow$  😊

## Validation: K-nearest neighbors

- $k = 1$  is a more flexible model, it gives perfect predictions on the training set, but does not generalize well to never before seen data
- $k = n$  is a more generalized model, it gives poor predictions on the training set but generalizes better to never before seen data.
- Optimal: somewhere in between.







# Exercises!



# Thank you for your attention!

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