

# Recap

# Families of ML algorithms

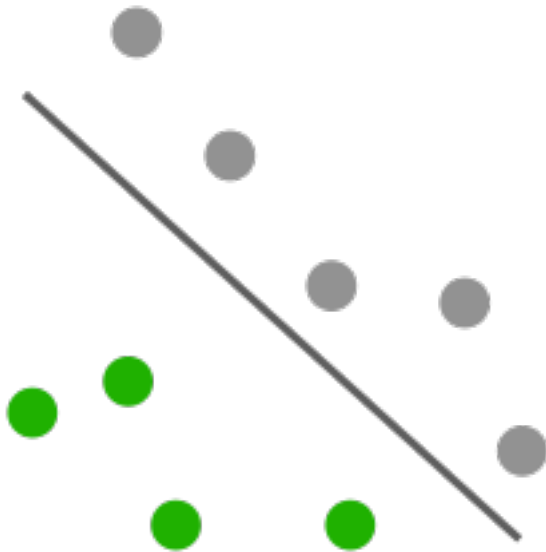
- Linear
- Tree-based
- kNN
- Neural Networks

# Linear model

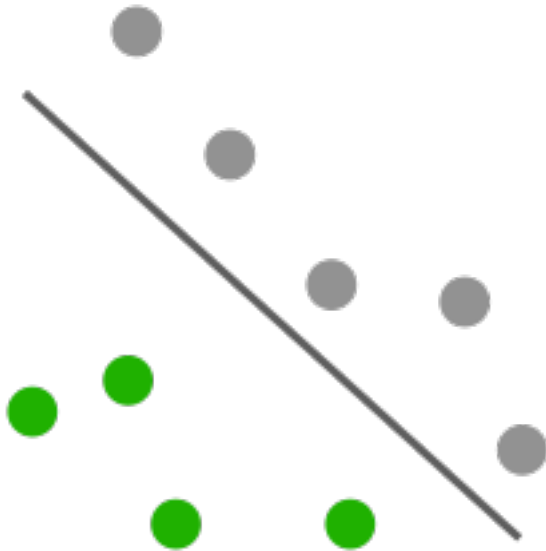
# Linear model



# Linear model



# Linear model

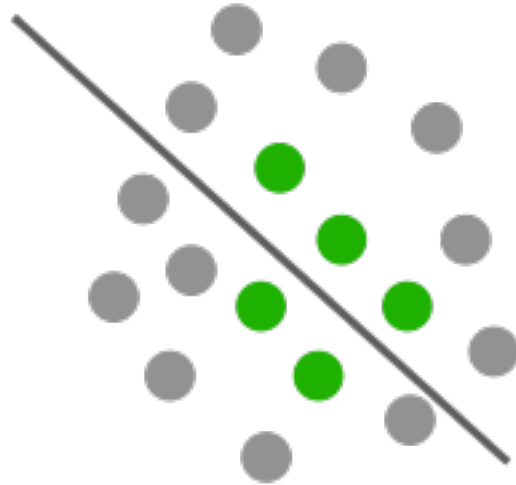


Examples:

- Logistic Regression
- Support Vector Machines

# Linear model

I want to emphasize that Linear Models are especially good for sparse high dimensional data. But you should keep in mind the limitations of Linear Models.



As an example, you can imagine two sets of points that form rings, one inside the other. Although it's pretty obvious how to separate them, Linear Models are not an appropriate choice either and will fail in this case.

# Linear model

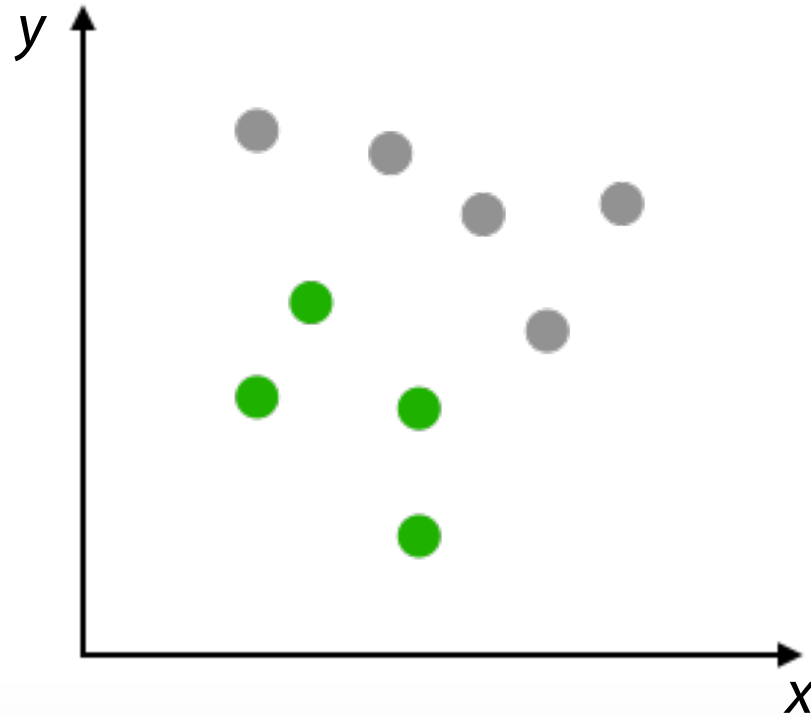




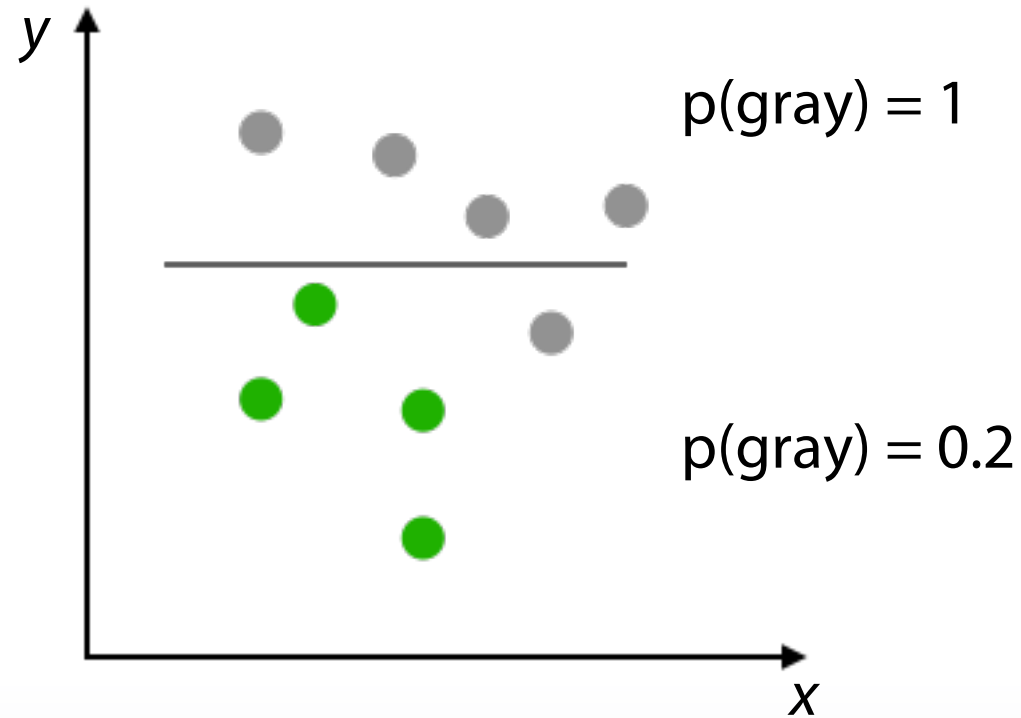
# Tree-based: Decision Tree, Random Forest, GBDT

Tree-Based Methods use decision tree as a basic block for building more complicated models.

# Tree-based: Decision Tree, Random Forest, GBDT

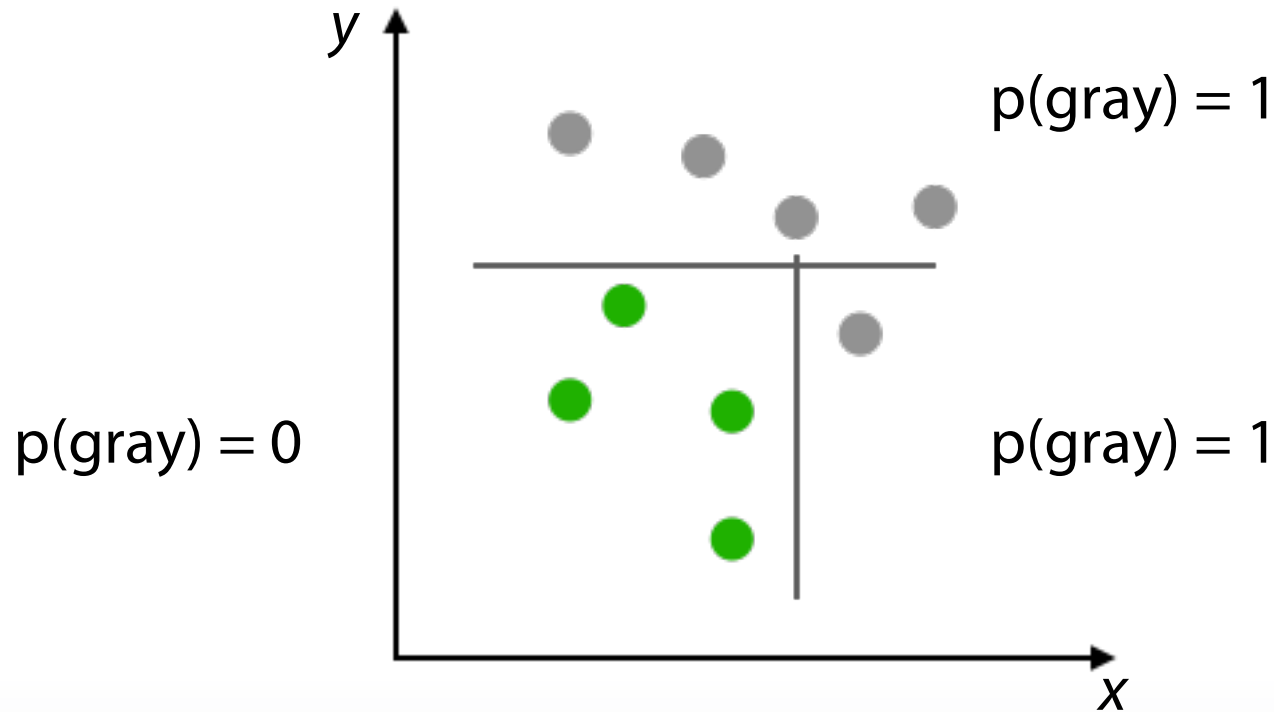


# Tree-based: Decision Tree, Random Forest, GBDT

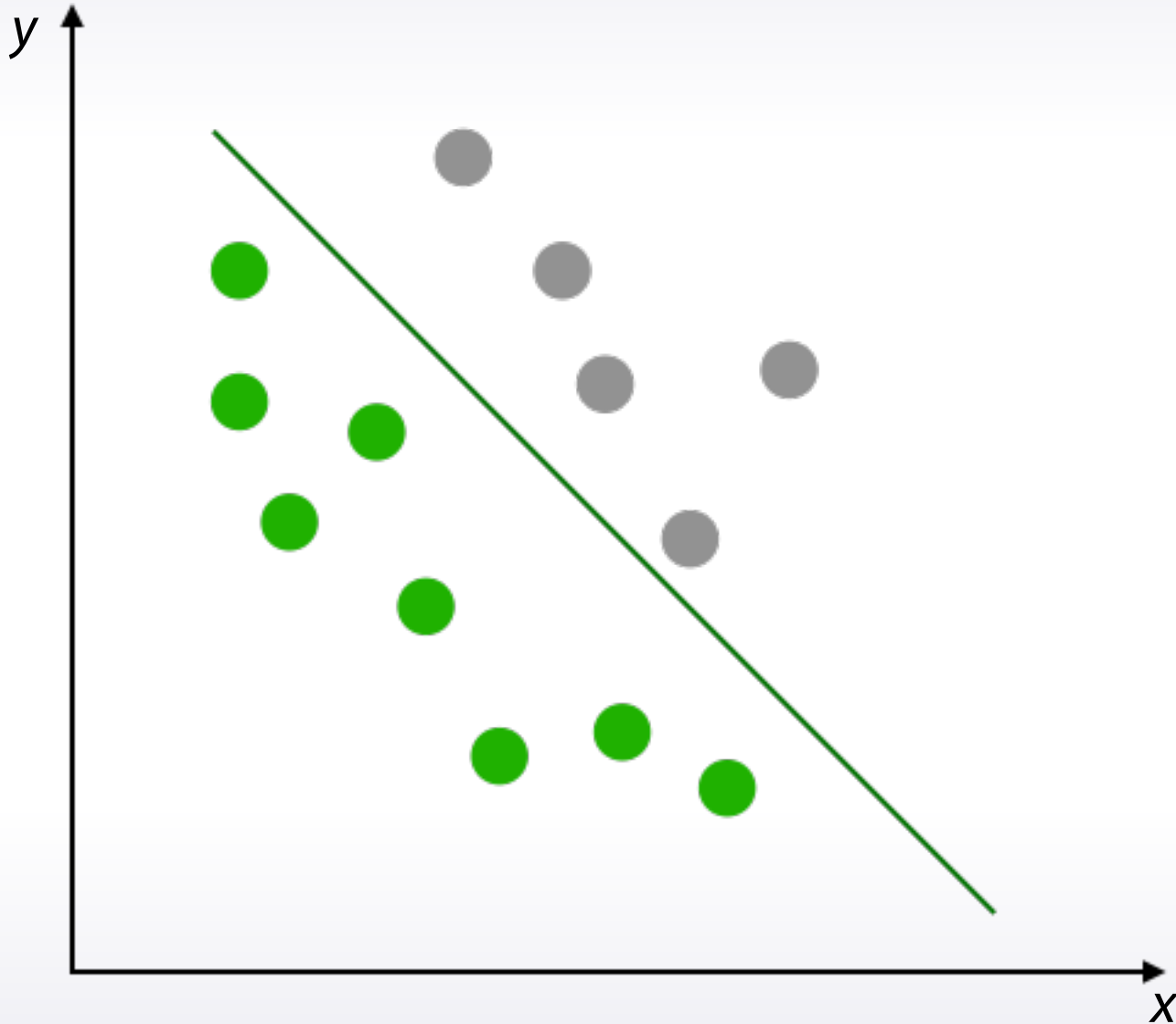


# Tree-based: Decision Tree, Random Forest, GBDT

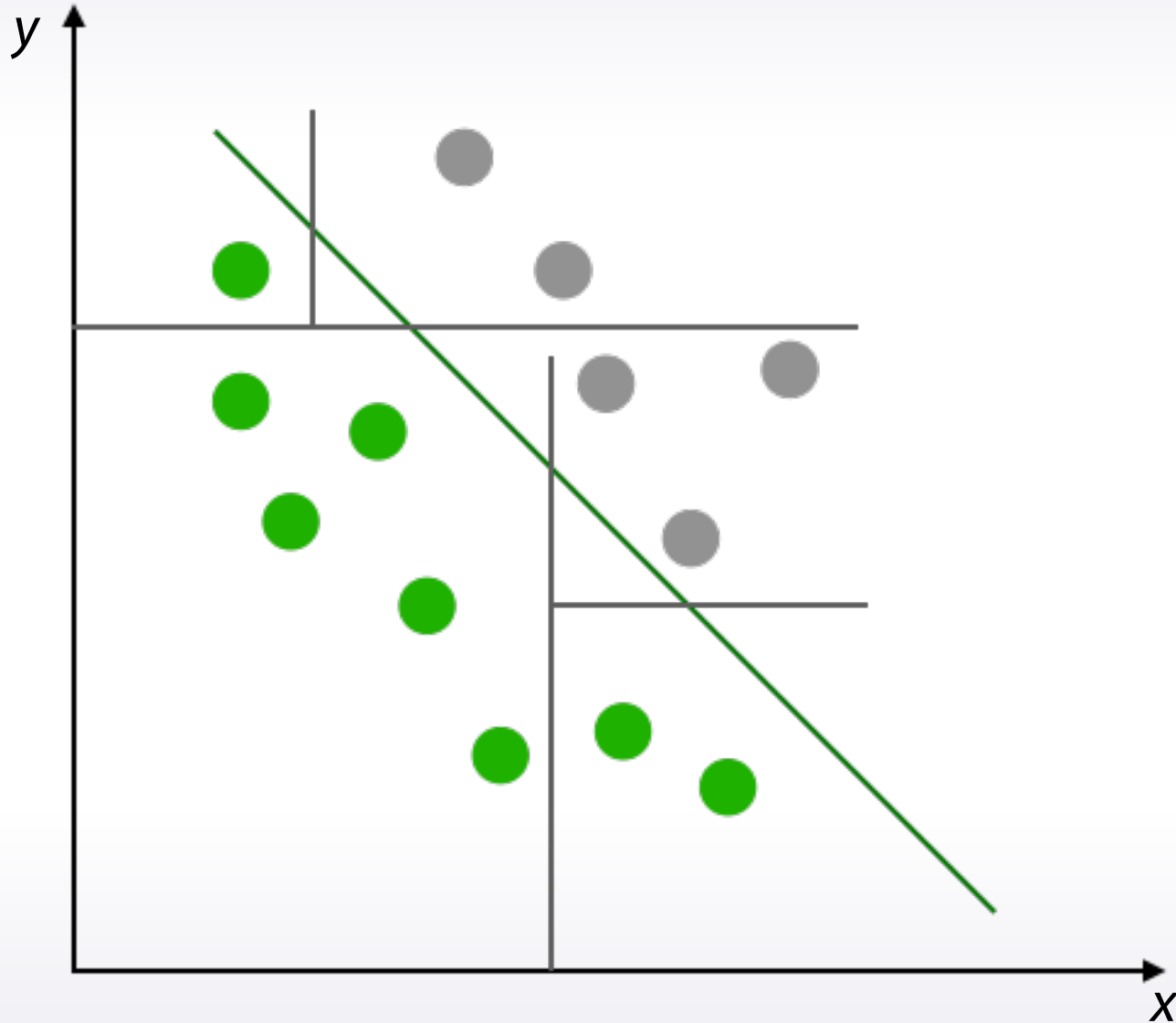
In general, tree-based models are very powerful and can be a good default method for tabular data. In almost every competitions, winners use this approach. But keep in mind that for Tree-Based Methods, it's hard to capture linear dependencies since it requires a lot of splits.



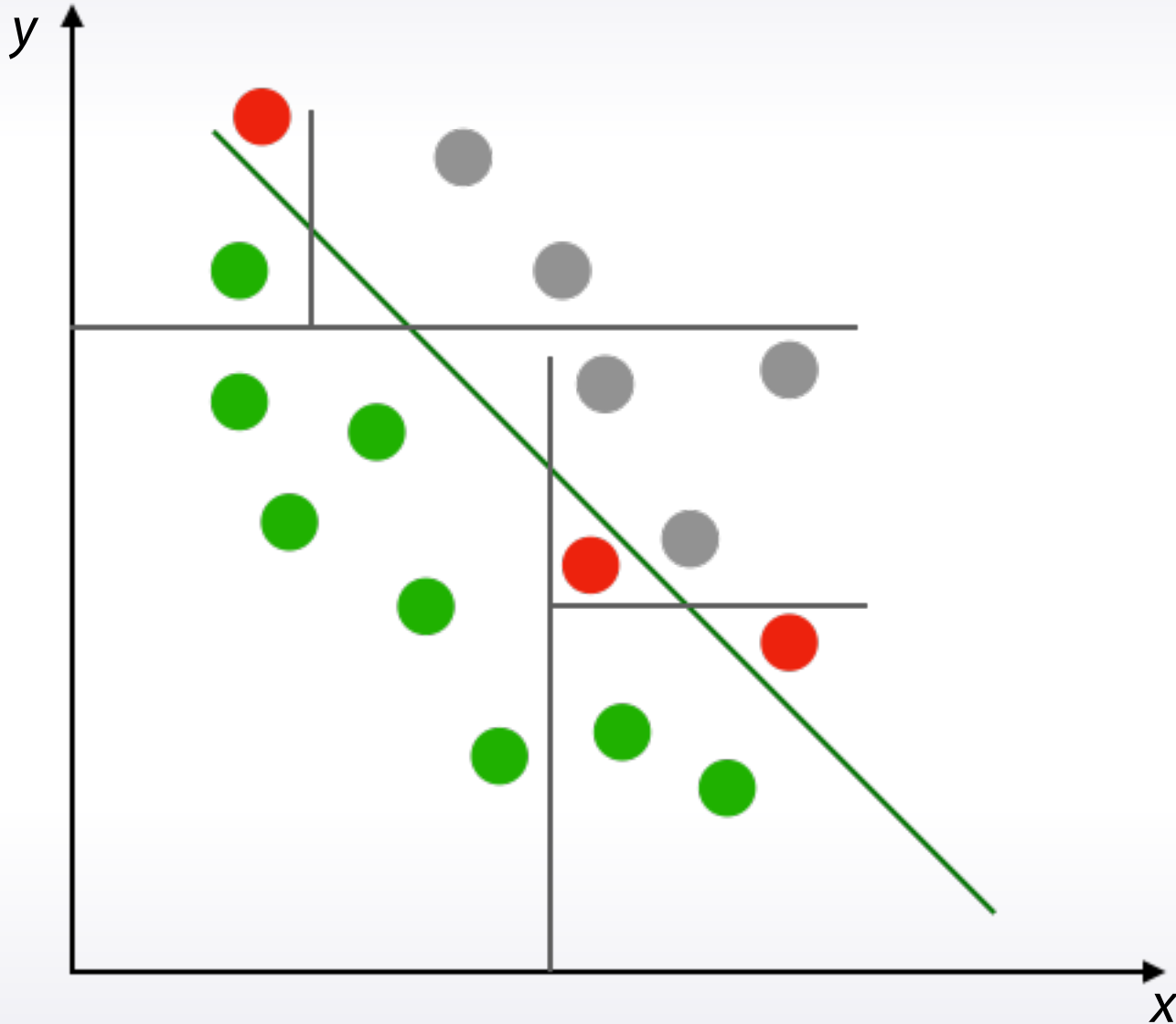
# Tree-based: Decision Tree, Random Forest, GBDT



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# Tree-based: Decision Tree, Random Forest, GBDT



# Tree-based methods



Scikit-Learn contains quite good implementation of random forest which I personally prefer. All the Scikit-Learn contain implementation of gradient boost decision trees. I prefer to use libraries like XGBoost and LightGBM for their higher speed and accuracy.

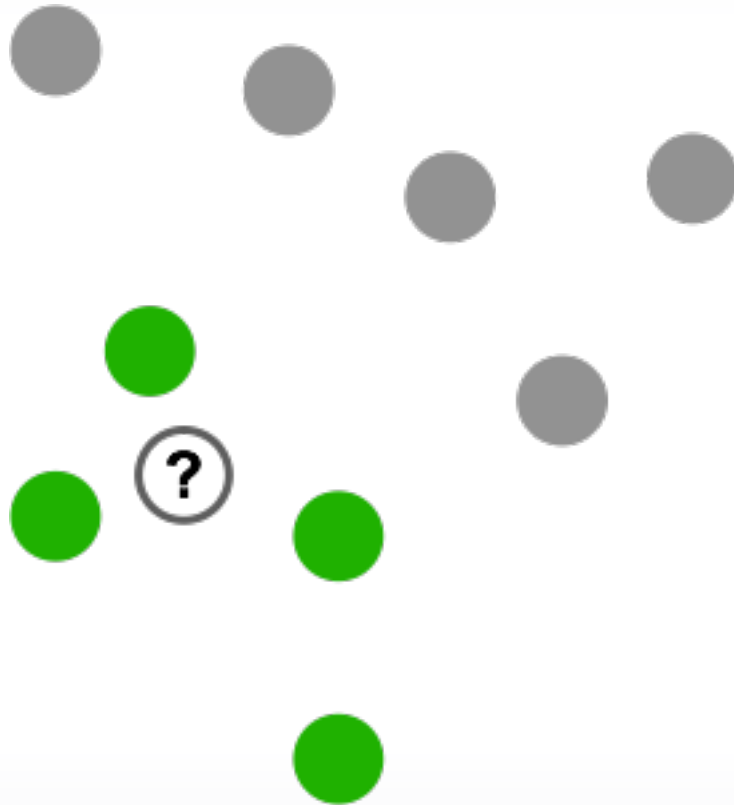


# kNN-based methods

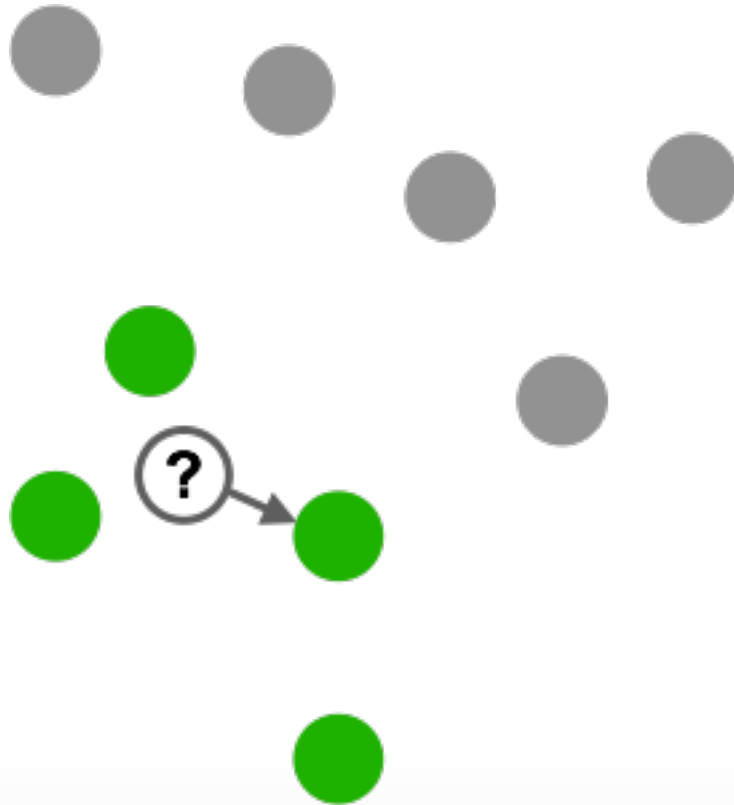
# kNN-based methods



# kNN-based methods



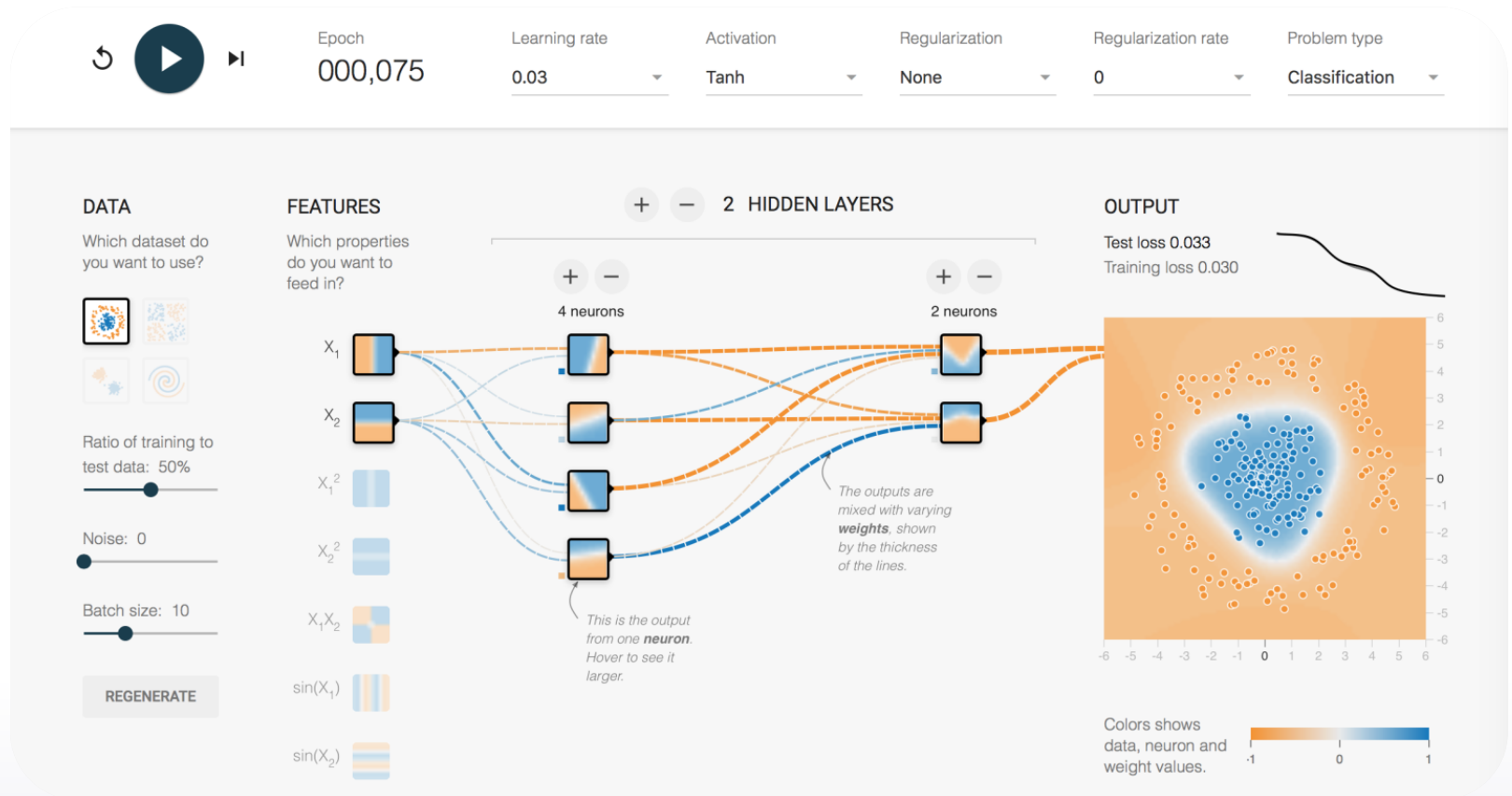
# kNN-based methods



# kNN-based methods



# Neural Networks



Tensorflow Playground, <http://playground.tensorflow.org>

# Neural Networks



*dmlc*  
***mxnet***

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Lasagne

# No Free Lunch Theorem



# No Free Lunch Theorem

“Here is no method which **outperforms all others**  
**for all tasks**”

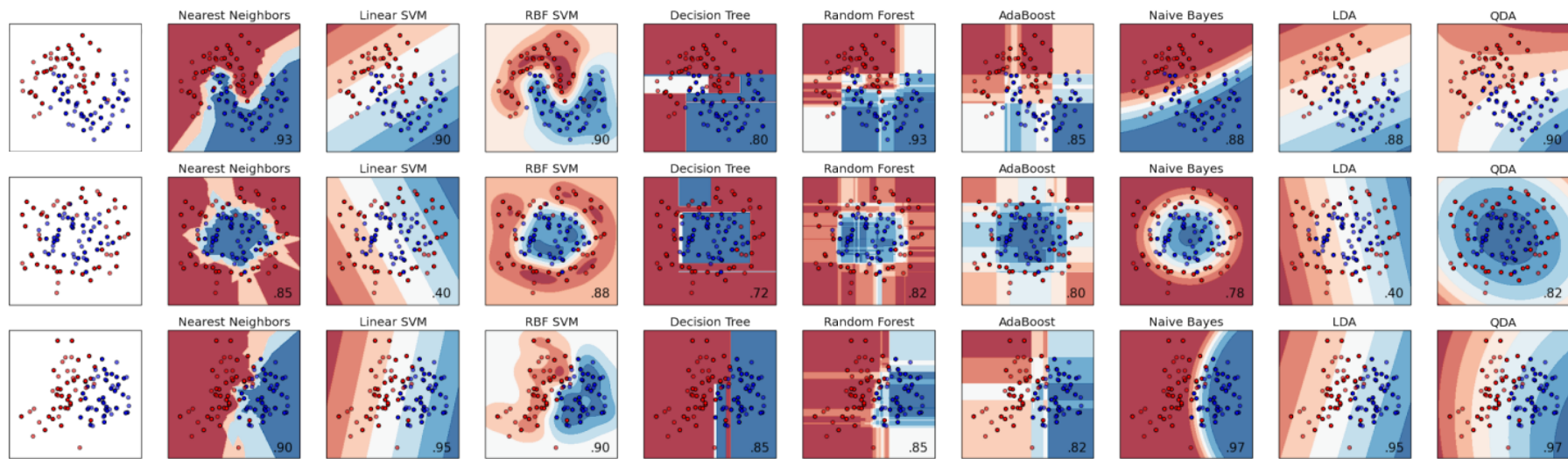
# No Free Lunch Theorem

“Here is no method which **outperforms all others**  
**for all tasks**”

or

“For every method **we can construct a task**  
for which **this particular method will not be the**  
**best**”

# Decision surfaces



Classifier comparison, [http://scikit-learn.org/stable/auto\\_examples/classification/plot\\_classifier\\_comparison.html](http://scikit-learn.org/stable/auto_examples/classification/plot_classifier_comparison.html)

# Conclusion

- There is no “silver bullet” algorithm
- Linear models split space into 2 subspaces
- Tree-based methods splits space into boxes
- k-NN methods heavy rely on how to measure points “closeness”
- Feed-forward NNs produce smooth non-linear decision boundary

The most powerful methods are  
**Gradient Boosted Decision Trees** and **Neural Networks**.  
But you shouldn't underestimate the others