

Algorithm	Best Fit Situation	Advantage	Disadvantage
Random Forest	<ol style="list-style-type: none"><li>1. Suited for at almost any machine learning problem</li><li>2. Ex. Bioinformatics</li></ol>	<ol style="list-style-type: none"><li>1. Can work in parallel</li><li>2. Susceptible to overfits</li><li>3. Automatically handles missing values if you impute using a special number</li><li>4. No need to transform any variable</li><li>5. No need to tweak parameters</li></ol>	<ol style="list-style-type: none"><li>6. Difficult to interpret</li><li>7. Weaker on regression when estimating values at the extremities of the distribution of response values</li><li>8. Biased in multiclass problems toward more frequent classes</li></ol>
Gradient Boosting	<ol style="list-style-type: none"><li>1. Apt at almost any machine learning problem</li><li>2. Search engines (solving the problem of learning to rank)</li></ol>	<ol style="list-style-type: none"><li>3. It can approximate most nonlinear function</li><li>4. Best in class predictor</li><li>5. Automatically handles missing values</li><li>6. No need to transform any variable</li></ol>	<ol style="list-style-type: none"><li>7. It can overfit if run for too many iterations</li><li>8. Sensitive to noisy data and outliers</li><li>9. Doesn't work at its best without parameter tuning</li></ol>
Linear Regression	<ol style="list-style-type: none"><li>1. Baseline prediction</li><li>2. Econometric predictions</li></ol>	<ol style="list-style-type: none"><li>1. Simple to understand and explain</li></ol>	<ol style="list-style-type: none"><li>1. You have to work hard to make it fit</li></ol>

	<ol style="list-style-type: none"><li>3. Modelling</li><li>4. marketing responses</li></ol>	<p>It seldom overfits</p> <ol style="list-style-type: none"><li>2. Using L1 &amp; L2 regularization is effective in feature selection</li><li>3. Fast to train</li><li>4. Easy to train on big data thanks to its stochastic version</li></ol>	<p>nonlinear functions</p> <ol style="list-style-type: none"><li>2. Can suffer from outliers</li></ol>
Support Vector Machine	<ol style="list-style-type: none"><li>1. Character recognition</li><li>2. Image recognition</li><li>3. Text classification</li></ol>	<ol style="list-style-type: none"><li>1. Automatic non-linear feature creation</li><li>2. Can approximate complex non-linear functions</li><li>3. Works only with a portion of the examples (the support vectors)</li></ol>	<ol style="list-style-type: none"><li>1. Difficult to interpret when applying non-linear kernels</li><li>2. Suffers from too many examples, after 10,000 examples it starts taking too long to train</li></ol>
K Nearest Neighbor	<ol style="list-style-type: none"><li>1. Computer vision</li><li>2. Multilabel tagging</li><li>3. Recommender systems</li><li>4. Spell checking problems</li></ol>	<ol style="list-style-type: none"><li>1. Fast, lazy training</li><li>2. Can naturally handle extreme multiclass problems (like tagging text)</li></ol>	<ol style="list-style-type: none"><li>1. Slow and cumbersome in the predicting phase</li><li>2. Can fail to predict correctly due to the curse of dimensionality</li></ol>

Adaboost	Facedetection	<ol style="list-style-type: none"><li>1. Automatically handles missing values</li><li>2. No need to transform any variable</li><li>3. It doesn't overfit easily</li><li>4. Few parameters to tweak</li><li>5. It can leverage many different weak-learners</li></ol>	<ol style="list-style-type: none"><li>1. Sensitive to noisy data and outliers</li><li>2. Never the best in class predictions</li></ol>
Naive Bayes	<ol style="list-style-type: none"><li>1. Face recognition</li><li>2. Sentiment analysis</li><li>3. Spam detection</li><li>4. Text classification</li></ol>	<ol style="list-style-type: none"><li>1. Easy and fast to implement, doesn't require too much memory and can be used for online learning</li><li>2. Easy to understand</li><li>3. Takes into account prior knowledge</li></ol>	<ol style="list-style-type: none"><li>1. Strong and unrealistic feature independence assumptions</li><li>2. Fails estimating rare occurrences</li><li>3. Suffers from irrelevant features</li></ol>
Neural Networks	<ol style="list-style-type: none"><li>1. Image recognition</li><li>2. Language recognition and translation</li><li>3. Speech recognition</li></ol>	<ol style="list-style-type: none"><li>1. It can approximate any non-linear function</li></ol>	<ol style="list-style-type: none"><li>1. It requires you to define a network architecture</li></ol>

	<ol style="list-style-type: none"><li>4. Vision recognition</li></ol>	<ol style="list-style-type: none"><li>2. Robust to outliers</li><li>3. It can work with image, text and sound data</li></ol>	<ol style="list-style-type: none"><li>2. Difficult to tune because of too many parameters and you have also to decide the architecture of the network</li><li>3. Difficult to interpret</li><li>4. Easy to overfit</li></ol>
Logistic Regression	<ol style="list-style-type: none"><li>1. Ordering results by probability</li><li>2. Modelling marketing responses</li></ol>	<ol style="list-style-type: none"><li>1. Simple to understand and explain</li><li>2. It seldom overfits</li><li>3. Using L1 &amp; L2 regularization is effective in feature selection</li><li>4. The best algorithm for predicting probabilities of an event</li><li>5. Fast to train</li><li>6. Easy to train on big data thanks to its stochastic version</li></ol>	<ol style="list-style-type: none"><li>1. You have to work hard to make it fit non-linear functions</li><li>2. Can suffer from outliers</li></ol>

SVD	Recommendation System	<ol style="list-style-type: none"><li>1. Can restructure data in a meaningful way</li></ol>	<ol style="list-style-type: none"><li>1. Difficult to understand why data has been restructured in a certain way</li></ol>
PCA	<ol style="list-style-type: none"><li>1. Removing collinearity</li><li>2. Reducing dimensions of the dataset</li></ol>	<ol style="list-style-type: none"><li>1. Can reduce data dimensionality</li></ol>	<ol style="list-style-type: none"><li>1. Implies strong linear assumptions (components are a weighted summations of features)</li></ol>
K means	<ol style="list-style-type: none"><li>1. Segmentation</li></ol>	<ol style="list-style-type: none"><li>1. Fast in finding clusters</li><li>2. Can detect outliers in multiple dimensions</li></ol>	<ol style="list-style-type: none"><li>1. Suffers from multicollinearit</li><li>2. Clusters are spherical, can't detect groups of other shape</li><li>3. Unstable solutions, depends on initialization</li></ol>