

# hyperparameters

Sunday, September 30, 2018 11:27 AM

Model	Parameters to optimize	Good range of values
Linear Regression	<ul style="list-style-type: none"><li>fit_intercept</li><li>normalize</li></ul>	<ul style="list-style-type: none"><li>True / False</li><li>True / False</li></ul>
Ridge	<ul style="list-style-type: none"><li>alpha</li><li>Fit_intercept</li><li>Normalize</li></ul>	<ul style="list-style-type: none"><li>0.01, 0.1, 1.0, 10, 100</li><li>True/False</li><li>True/False</li></ul>
k-neighbors	<ul style="list-style-type: none"><li>N_neighbors</li><li>p</li></ul>	<ul style="list-style-type: none"><li>2, 4, 8, 16 ....</li><li>2, 3</li></ul>
SVM	<ul style="list-style-type: none"><li>C</li><li>Gamma</li><li>class_weight</li></ul>	<ul style="list-style-type: none"><li>0.001, 0.01....10...100...1000</li><li>'Auto', RS*</li><li>'Balanced' , None</li></ul>
Logistic Regression	<ul style="list-style-type: none"><li>Penalty</li><li>C</li></ul>	<ul style="list-style-type: none"><li>L1 or L2</li><li>0.001, 0.01....10...100</li></ul>
Naive Bayes (all variations)	NONE	NONE
Lasso	<ul style="list-style-type: none"><li>Alpha</li><li>Normalize</li></ul>	<ul style="list-style-type: none"><li>0.1, 1.0, 10</li><li>True/False</li></ul>
Random Forest	<ul style="list-style-type: none"><li>N_estimators</li><li>Max_depth</li><li>Min_samples_split</li><li>Min_samples_leaf</li><li>Max_features</li></ul>	<ul style="list-style-type: none"><li>120, 300, 500, 800, 1200</li><li>5, 8, 15, 25, 30, None</li><li>1, 2, 5, 10, 15, 100</li><li>1, 2, 5, 10</li><li>Log2, sqrt, None</li></ul>

Xgboost	<ul style="list-style-type: none"><li>Eta</li><li>Gamma</li><li>Max_depth</li><li>Min_child_weight</li><li>Subsample</li><li>Colsample_bytree</li><li>Lambda</li><li>alpha</li></ul>	<ul style="list-style-type: none"><li>0.01,0.015, 0.025, 0.05, 0.1</li><li>0.05-0.1,0.3,0.5,0.7,0.9,1.0</li><li>3, 5, 7, 9, 12, 15, 17, 25</li><li>1, 3, 5, 7</li><li>0.6, 0.7, 0.8, 0.9, 1.0</li><li>0.6, 0.7, 0.8, 0.9, 1.0</li><li>0.01-0.1, 1.0 , RS*</li><li>0, 0.1, 0.5, 1.0 RS*</li></ul>
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## XG boost hyperparameters

Depth of tree:--imp as it results in complexity. 5 is better not more  
Learning rate:--shrinkage parameter..learning the right weights  
Max\_delta\_step:(xg boost handles one vs all which is beneficial, max\_delta\_step helps in imbalanced data sets)  
No of estimators :- no of boosting rounds)  
Booster--mostly tree based used  
Base\_score :--> not needed in parameterization  
Random\_state ==>  
Missing (specify if missing denotes some value)

Sample size  
No of columns  
Regularization params :-  
lamda (L2 regularizn)  
Alpha L1 regularizn  
Gamma L0 regularizn  
Gamma L0  
Leads weights  
Min\_child weight :--bases on which a tree can be made

## Random Forest

n\_estimators = number of trees in the forest

- max\_features = max number of features considered for splitting a node
- max\_depth = max number of levels in each decision tree
- min\_samples\_split = min number of data points placed in a node before the node is split
- min\_samples\_leaf = min number of data points allowed in a leaf node
- bootstrap = method for sampling data points (with or without replacement)

From <<https://towardsdatascience.com/hyperparameter-tuning-the-random-forest-in-python-using-scikit-learn-28d2aa77dd74>>