| Model | Task | Useful For | Equation | Looks Like |
|-----------------------------------------------------------|-------------------------------------|-----------------------------------------------------------------------------------------------|-----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|--------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|
| Linear Regression Least Squares LASSO Ridge Regression | Regression | Inference and Prediction (for LASSO and Ridge, Bootstrap or Bayesian approaches to inference) | $\hat{f}(x_i) = \beta_0 + \beta_1 X_{i1} + \dots + \beta_p X_{ip}$ | X_1 |
| Logistic Regression | Classification | Inference and Prediction | $\hat{f}_1(x_i) = \frac{e^{\beta_0 + \beta_1 X_{i1} + \dots + \beta_p X_{ip}}}{1 + e^{\beta_0 + \beta_1 X_{i1} + \dots + \beta_p X_{ip}}}$ | f(X1, X2) X1 The state of the |
| KNN | Regression and Classification | Prediction | $\hat{f}(x_i) = \frac{1}{K} \sum_{i \in N_0} Y_i$ $\hat{f}_j(x_i) = \frac{1}{K} \sum_{i \in N_0} \mathbb{I}_{\{j\}}(Y_i)$ | be zero de |
| Classification and Regression Trees | Regression and Classification | Prediction | $\hat{f}(x_i) = \sum_{m=1}^{ T } I_{R_m}(x_i) \hat{y}_m$ $\hat{y}_m is the mean training set response in region m for regression, or the proportion of training set responses in class j from region m.$ | $ \hat{y} = 1 $ $ \hat{y} = 2 $ $ \hat{y} = 3 $ $ \hat{y} = 4 $ |

| Ensembling Approach | Useful For | Component Models Differentiated By | Timing of Component Model Estimation | Component Models Combined Via | | | | |
|-----------------------------------------------------------------------|-------------------------------------------------------------------------------|-----------------------------------------------------------------------------------------------------------------------------------------|------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|------------------------------------------------------------------------------------------------------------------------------------------|--|--|--|--|
| Stacking | Could be anything; often, different types of models (linear model, KNN, CART) | | Separate Estimation of Each Component Model, Combined Later | Stage 2 model takes predictions from stage 1 models as inputs. (Stage 2 could be a simple average or majority vote, or a complex model). | | | | |
| Boosting | Prediction | Each component model trained on residuals from previous models, or more weight assigned to cases not predicted well by previous models. | Sequential Estimation of Component Models, Estimation of Each Model Depends on Previous Models | Typically, additive (predictions from component models are added together) | | | | |
| Approach for Model Differentiation Component Models Differentiated By | | | | | | | | |
| | Bagging | | Models trained on bootstrap resampled data sets | | | | | |
| | Feature Subsets | Models | Models use different subsets of features, or use features differently in estimation process | | | | | |
| | | | | | | | | |
| Named Ensemble Model | | | Component Models Differentiated By | | | | | |
| | Random Forest | tree is tra | All component models are trees. Combines bagging and feature subsets: each tree is trained on a bagged data set, and a randomly selected subset of features are used in finding each split as the tree is grown. | | | | | |