$$|h_1| \Rightarrow y_1$$

$$|h_2| \Rightarrow y_2$$

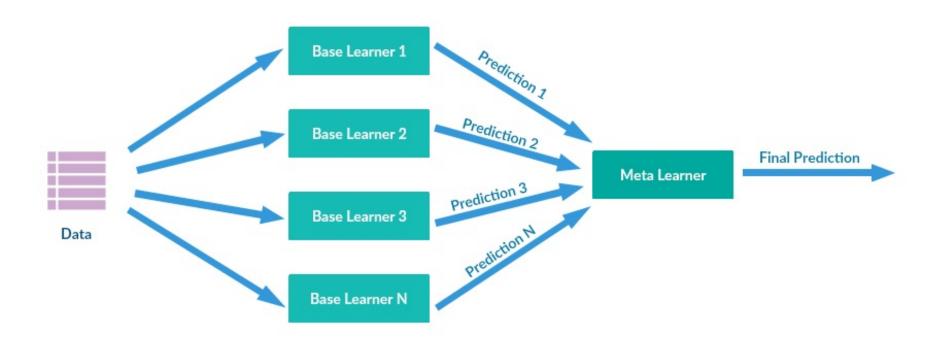
$$|h_7| \Rightarrow y_7$$

$$= sign\left(\sum_{t\geq 1}^{\infty} h_t(x)\right)$$

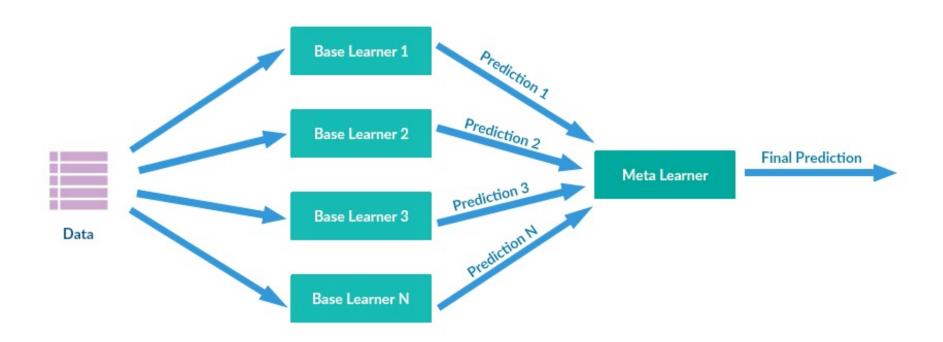
$$h_{1}(n) \in \{+1, -1\}$$
 \vdots
 $h_{T}(n) \in \{+1, -1\}$

Ensemble Learning

g weak Learner



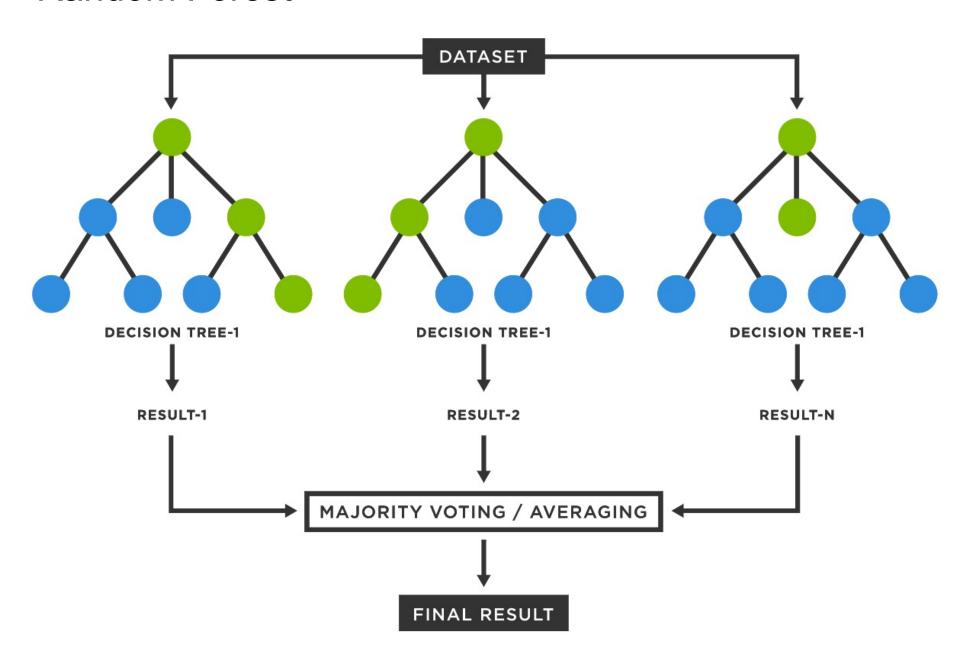
Ensemble Learning



Python: voting classifier

```
# Train Model
voting_classifier.fit(X_train, y_train)
```

Random Forest



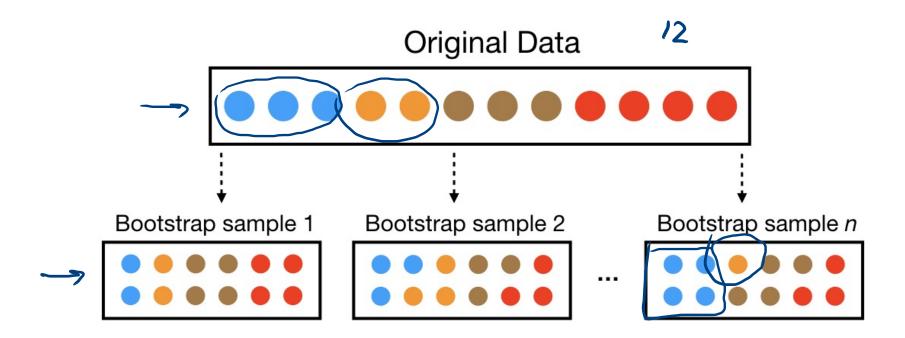
Ensemble Learning Bagging: Bootstrap + Aggregation

Boosting Adaptive Boosting

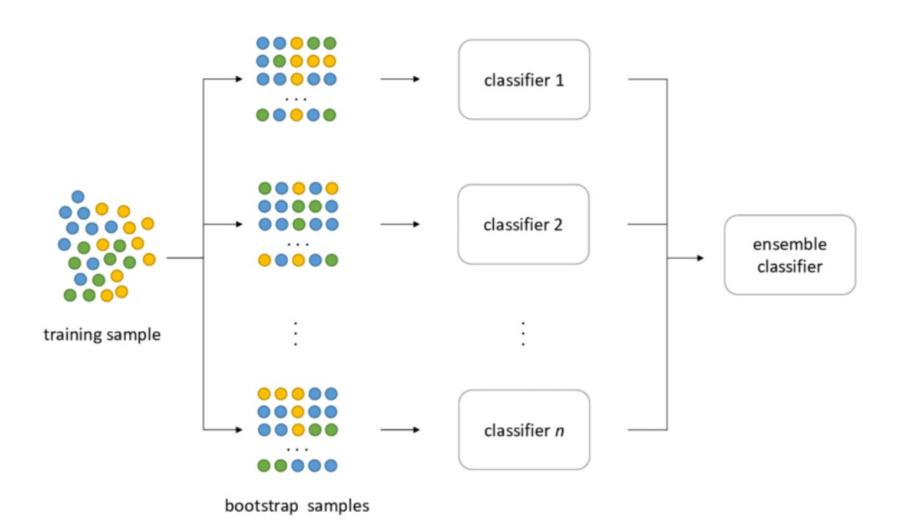
Ada Boost

Bootstrapping





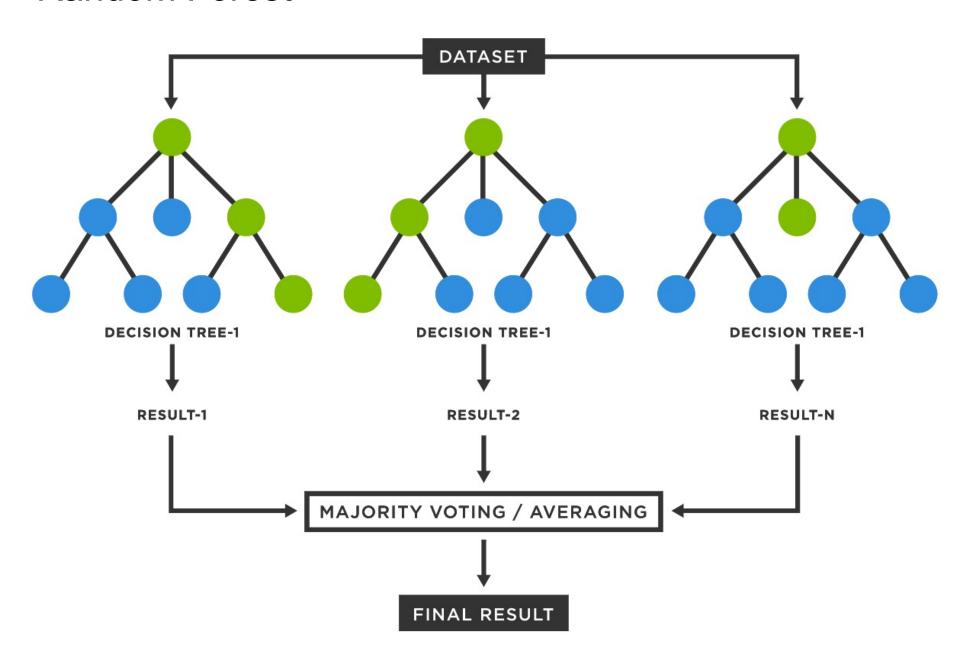
Bagging: Bootstrapping + Aggregation



Python: Bagging

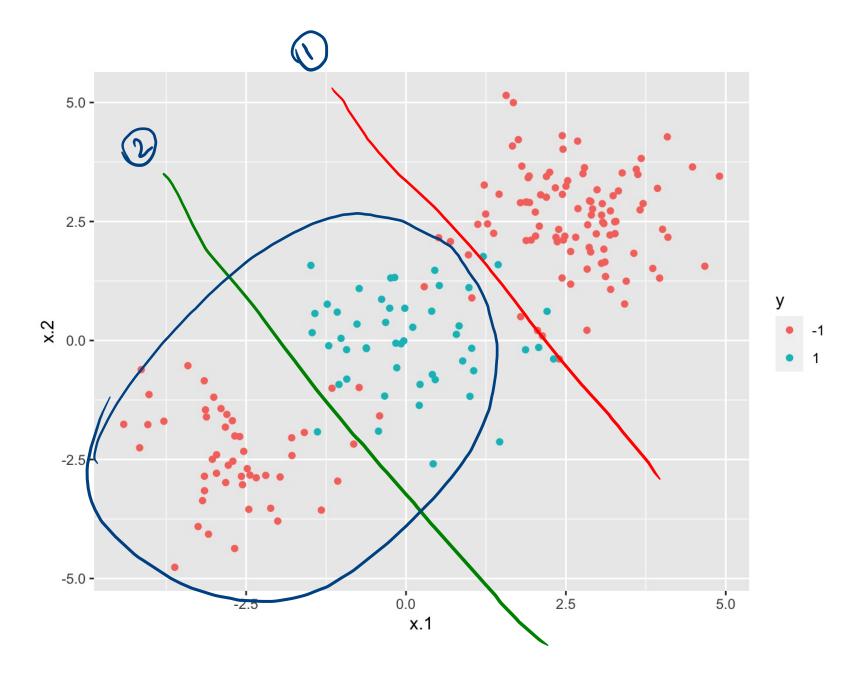
```
# Train the Model
bagging_classifier.fit(X_train, y_train)
```

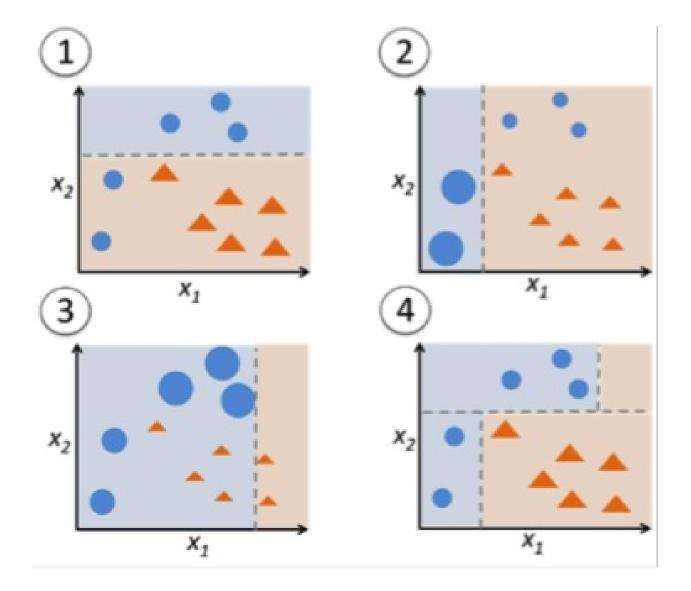
Random Forest

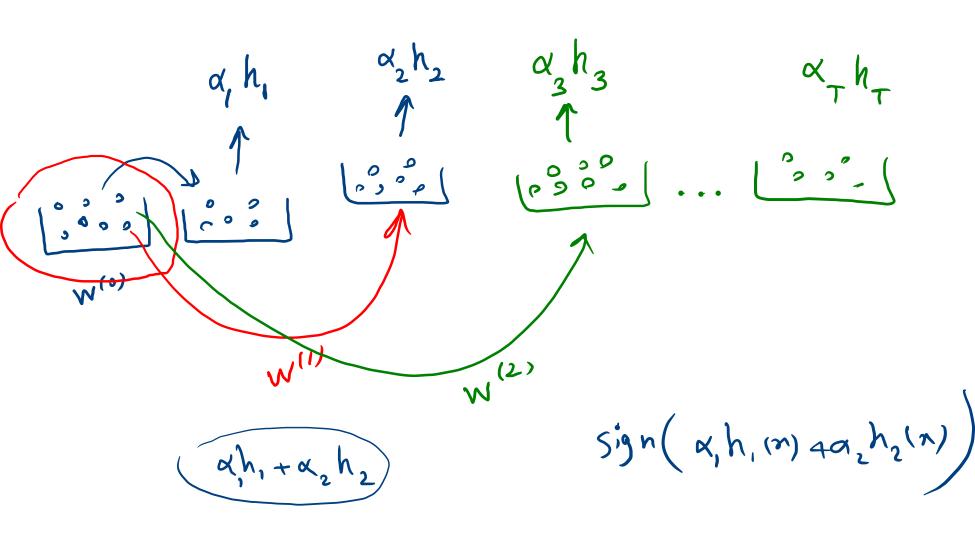


Python: Random Forest

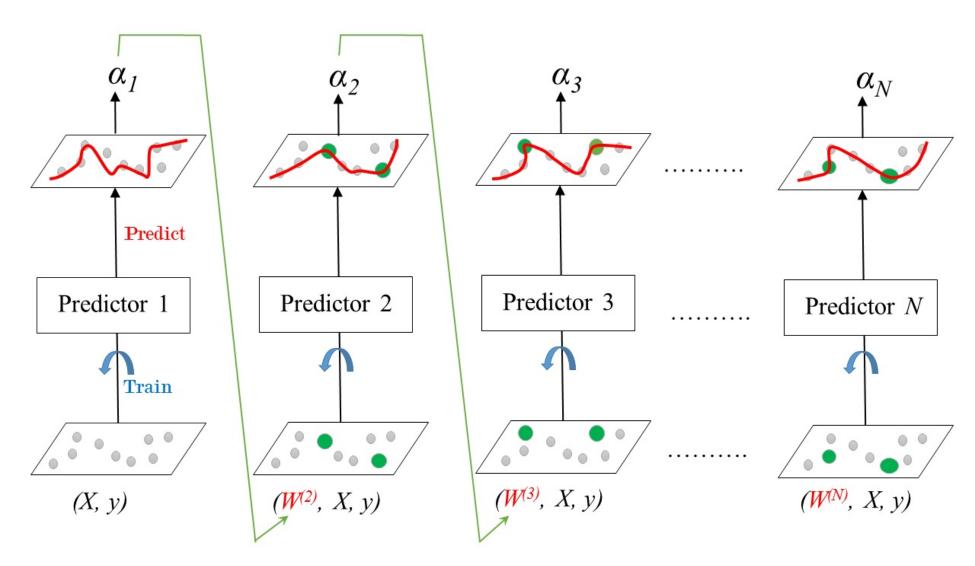
```
# Train the model
rf_model.fit(X_train, y_train)
```







AdaBoost



Python: AdaBoost

```
# Train Model
adaboost_model.fit(X_train, y_train)
```

$$E = \sum_{i=1}^{n} I(y, h(n_i) < 0)$$

$$E = \sum_{i=1}^{n} e - y, h(n_i)$$

$$= yh(n)$$

$$\begin{aligned}
F_{T,1}(x) &= \sum_{t=1}^{T} \alpha_{t} h_{t}(x) &= \sum_{t=1}^{T-1} \alpha_{t} h_{t}(x) + \alpha_{T} h_{T}(x) \\
E &= \sum_{i=1}^{n} e^{-y_{i} F_{T}(x_{i})} &= \sum_{i=1}^{n} e^{-y_{i} (F_{T,1}(x_{i}) + \alpha_{T} h_{T}(x_{i}))} \\
&= \sum_{i=1}^{n} e^{-y_{i} F_{T,1}(x_{i})} &= \sum_{i=1}^{n} W_{i}^{(T-1)} e^{x_{i} A_{T} h_{T}(x_{i})} \\
&= \sum_{i=1}^{n} e^{-y_{i} F_{T,1}(x_{i})} &= \sum_{i=1}^{n} W_{i}^{(T-1)} e^{x_{i} A_{T} h_{T}(x_{i})} \\
&= W_{i}^{(T-1)} V_{i}^{(T-1)} &= V_{i}^{(T-1)} e^{x_{i} A_{T} h_{T}(x_{i})}
\end{aligned}$$

$$W_{i}^{(T-i)} = \begin{array}{c} -\alpha_{T} \\ e \end{array} + \begin{array}{c} \sum_{i \in \mathcal{Y}_{i} \neq h_{T}(x_{i})} W_{i}^{(T-i)} e^{\alpha_{T}} \\ M \end{array}$$

$$= \frac{\sum_{i: y_i = h_T(n_i)} W_i^{(T-i)} e^{\alpha_T}}{\sum_{i \in N} W_i^{(T-i)}} + \frac{\sum_{i: y_i \neq h_T(n_i)} W_i^{(T-i)} e^{\alpha_T}}{\sum_{i \in N} W_i^{(T-i)}} = 0$$

$$\alpha_T = \frac{1}{2} \ln \frac{\sum_{i \in C} W_i^{(T-i)}}{\sum_{i \in N} W_i^{(T-i)}}$$

$$\epsilon_{T} = \frac{\sum_{i \in M} w_{i}^{(T-1)}}{\sum_{i=1}^{n} w_{i}^{(T-1)}}$$

$$T = \frac{i \in M}{i \in M}$$

$$\sum_{i=1}^{N} w_i^{(T-1)}$$

$$\sum_{i=1}^{N} w_i^{(T-1)}$$

$$\alpha_{\tau} = \frac{1}{2} \ln \frac{1 - \epsilon_{\tau}}{\epsilon_{\tau}}$$

$$E_{T} = 0 \longrightarrow \alpha_{T} = +\infty$$

$$E_{T} = 1 \longrightarrow \alpha_{T} = -\infty$$

$$\epsilon_{T} = \frac{1}{2} \longrightarrow \alpha_{T} = 0$$