

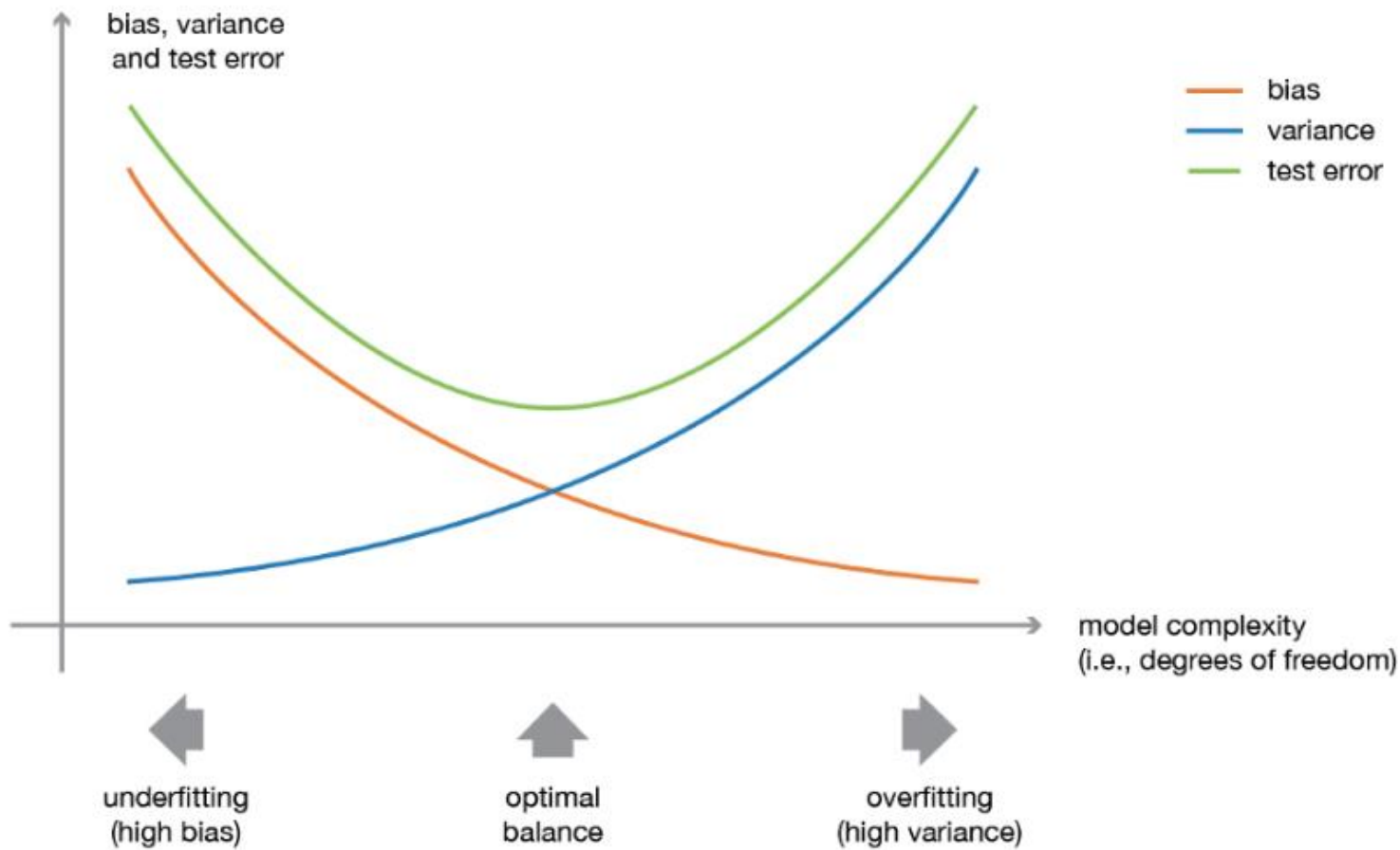
# Ensemble Learning

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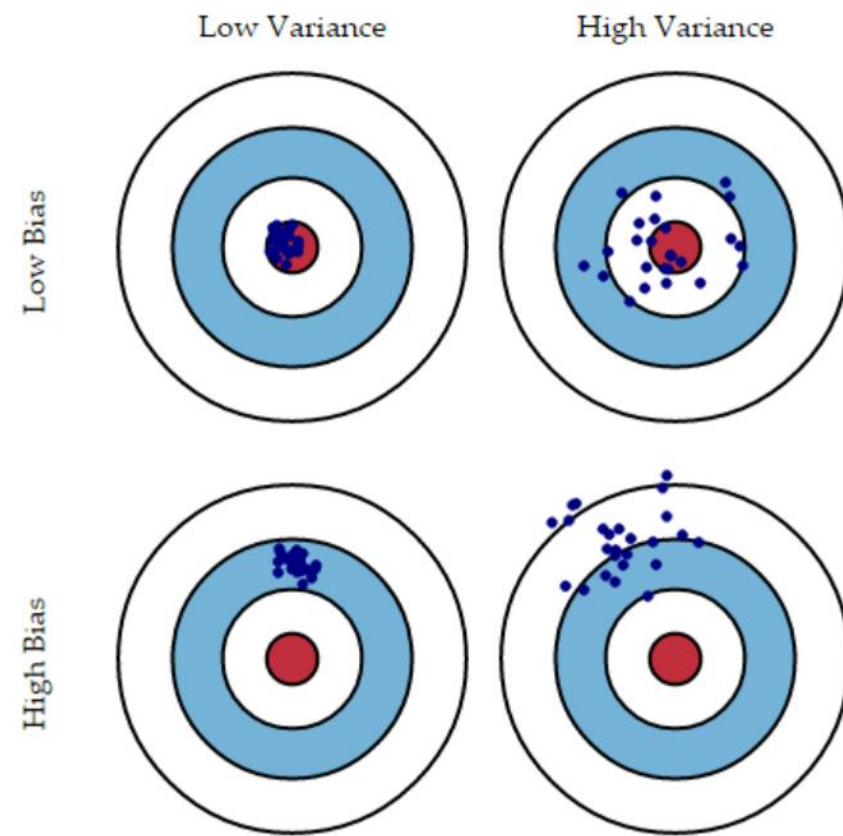
# Ensemble learning

Combine multiple weak algorithms to form a strong model.

Using ensemble methods allows to produce better predictions compared to a single model



# Bias Variance Trade off



# BAGGING

M1

M2

M3

M4

M5

Average / Majority vote

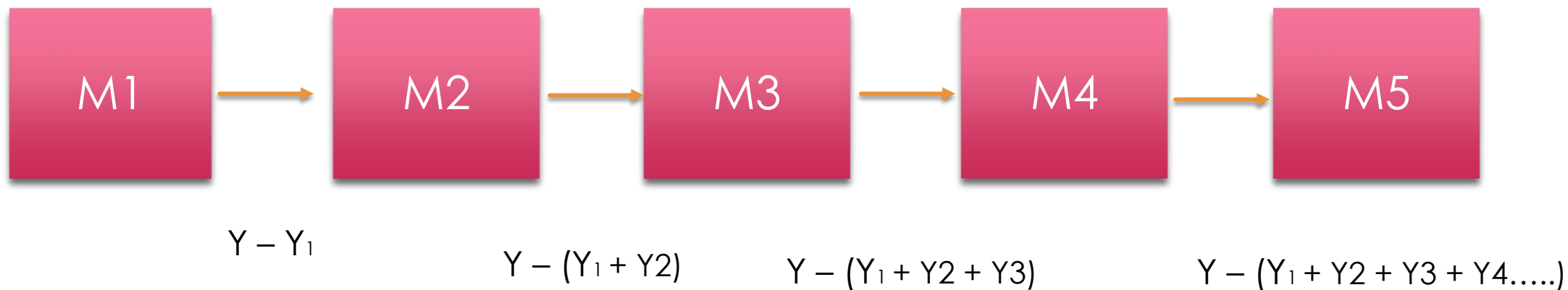
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graph TD; M1[M1] --> Avg[Average / Majority vote]; M2[M2] --> Avg; M3[M3] --> Avg; M4[M4] --> Avg; M5[M5] --> Avg;
```

The diagram illustrates the Bagging process. At the top, a dark purple banner contains the word 'BAGGING' in white. Below this, five pink rectangular boxes labeled M1, M2, M3, M4, and M5 are arranged horizontally. Five purple arrows point from each of these boxes down to a single orange rectangular box at the bottom labeled 'Average / Majority vote'. The entire diagram is framed by 'MG ANALYTICS' text on the left and right sides.

# BAGGING

- ▶ Bagging consists of building different parallel models
- ▶ Each model has different set of input samples helping to create unique models.
- ▶ Result is generated by taking the average of those predictions.
- ▶ This is useful when you want to decrease the variance while keeping the bias same.
- ▶ It works this way because bagging is kind of an averaging technique.
- ▶ Bagging helps when it is applied to an over fitted model by decreasing variance error.
- ▶ It does not help much with models which have high bias.

## BOOSTING

 $X, Y$ 

# BOOSTING

- ▶ Boosting consists of building different sequential models one after another.
- ▶ Each model has same  $X$  as input.
- ▶ First model predicts  $Y$ .
- ▶ then onwards models predict the error value left from previous model until the error is 0.
- ▶ **Decreases the bias error** and builds strong predictive models.
- ▶ They may sometimes over fit on the training data.
- ▶ For each iteration, boosting updates the weights of the samples, so that, samples that are misclassified by the ensemble can have a higher weight, and therefore, higher probability of being selected for training the new classifier.



- ▶ Bagging will mainly focus at getting an ensemble model with less variance than its components.
- ▶ Boosting and stacking will mainly try to produce strong models less biased than their components (even if variance can also be reduced).

# Stacking

STACKING ALLOWS TO CREATE A LINEAR COMBINATION OF MULTIPLE NON LINEAR MODELS.

Stacking creates a hierarchy of models using the outputs from previous layers.

# Base Models / Weak Learners

- ▶ Building blocks for designing more complex models.
- ▶ do not perform well :
  - ▶ high bias or too much variance.
- ▶ Ensemble methods try reduce bias and/or variance of such weak learners by combining several of them.
- ▶ Create a **strong learner** (or **ensemble model**) that achieves better performances.

Which model can be a WL?

Linear model  
with high  
penalty

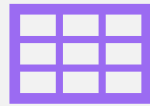
Linear model  
with subset  
of variables

Dtree stumps  
/ Shallow  
trees.

## Why to use WL?



Cannot learn niche patterns hence cannot overfit.



A combination of these will capture a general pattern.



As results would be combined will not be impacted by noise.