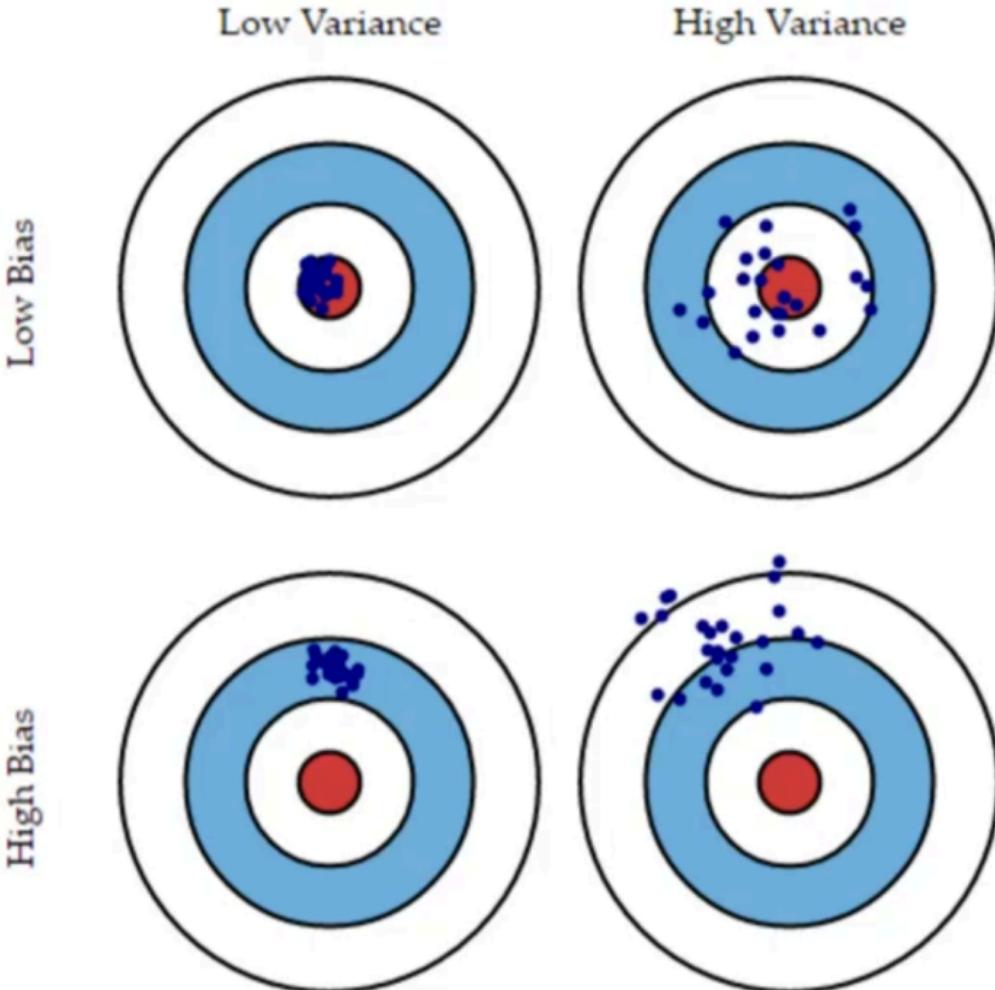


Bias Variance Trade Off is a fundamental topic of understanding your model's performance!

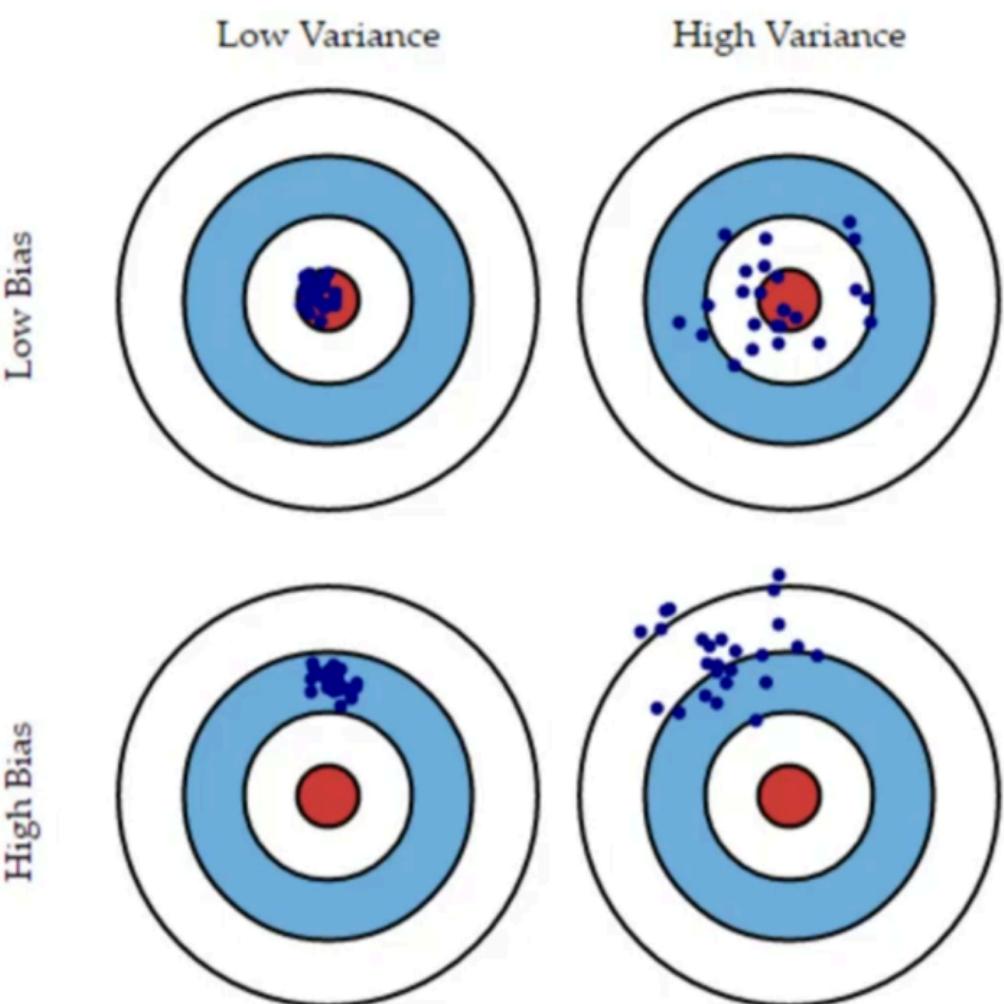
Review Chapter 2 of **Introduction to Statistical Learning** for a more in depth look!

- The bias-variance trade-off is the point where we are adding just noise by adding model complexity (flexibility).
- The training error goes down as it has to, but the test error is starting to go up.
- The model after the bias trade-off begins to overfit.

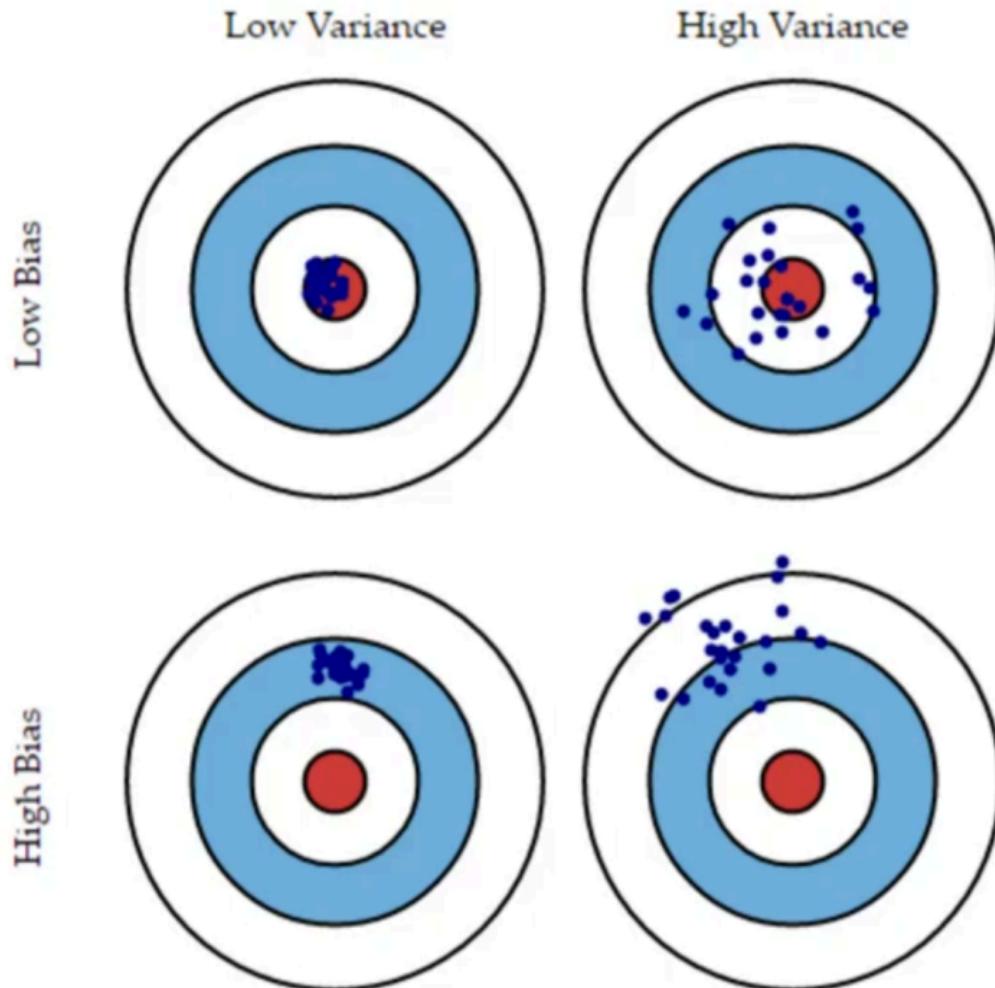
- Imagine that the center of the target is a model that perfectly predicts the correct values.
- As we move away from the bulls-eye, our predictions get worse and worse.



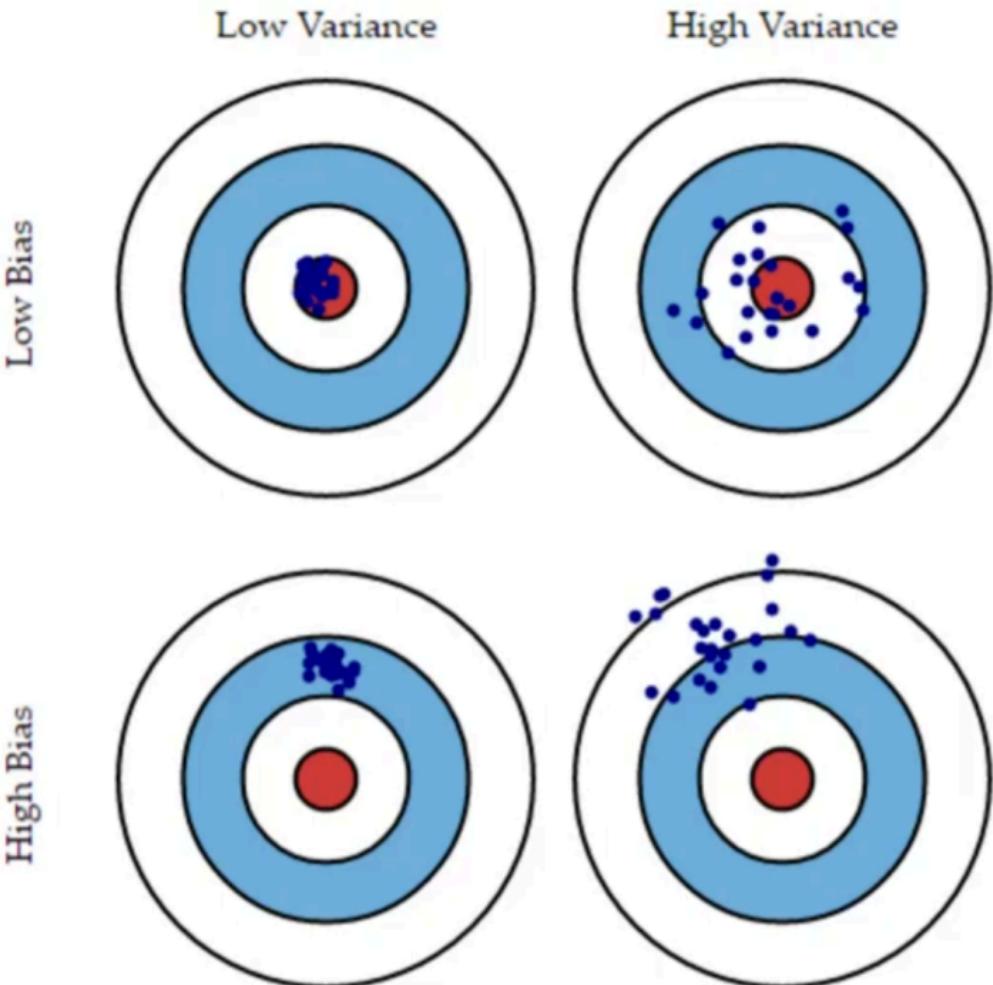
- Imagine we can repeat our entire model building process to get a number of separate hits on the target.
- Each hit represents an individual realization of our model, given the chance variability in the training data we gather.



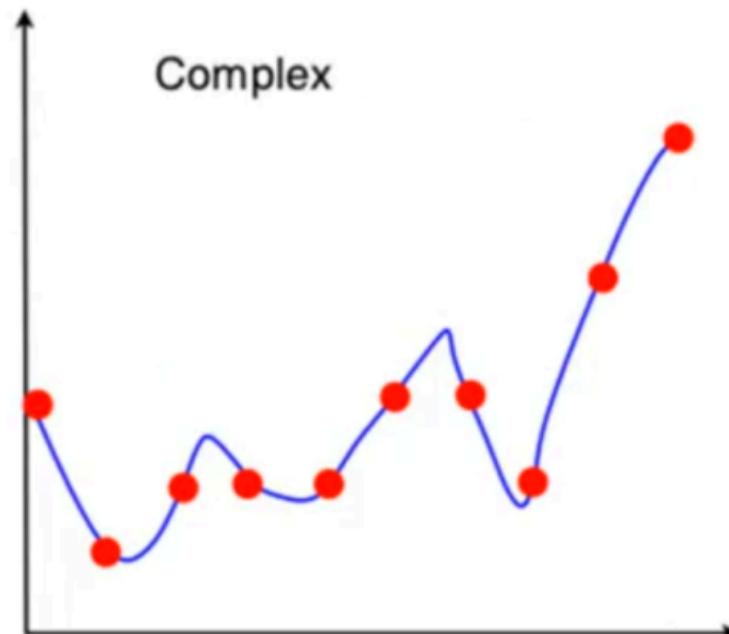
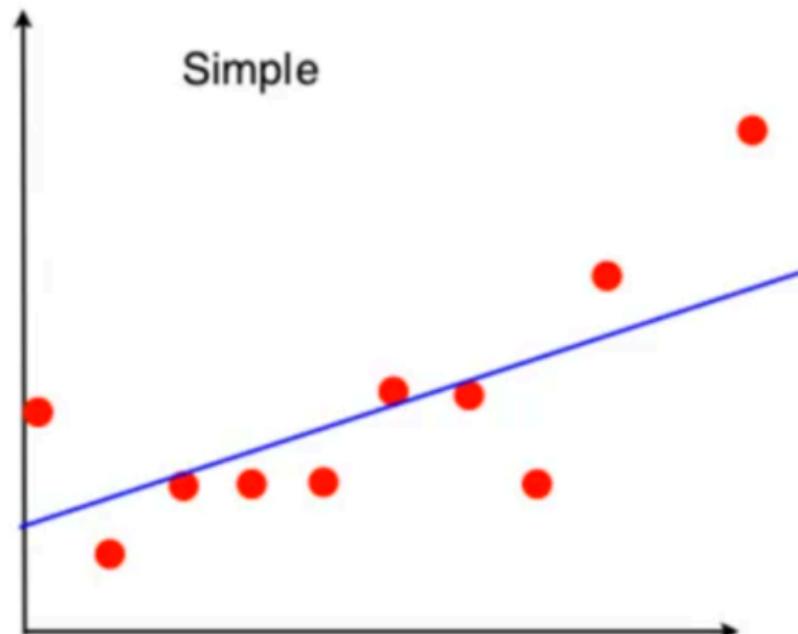
- Sometimes we will get a good distribution of training data so we predict very well and we are close to the bulls-eye, while sometimes our training data might be full of outliers or non-standard values resulting in poorer predictions.



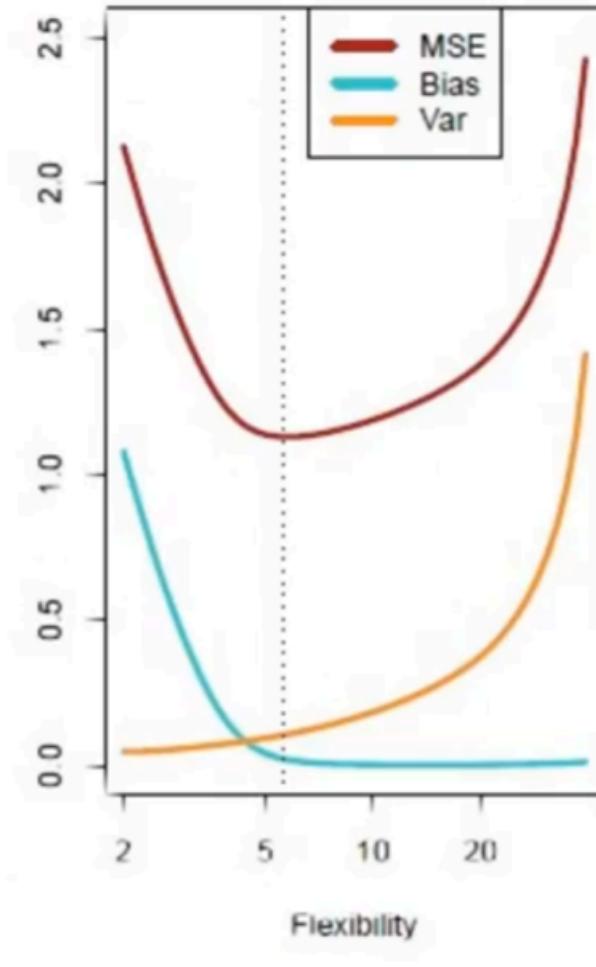
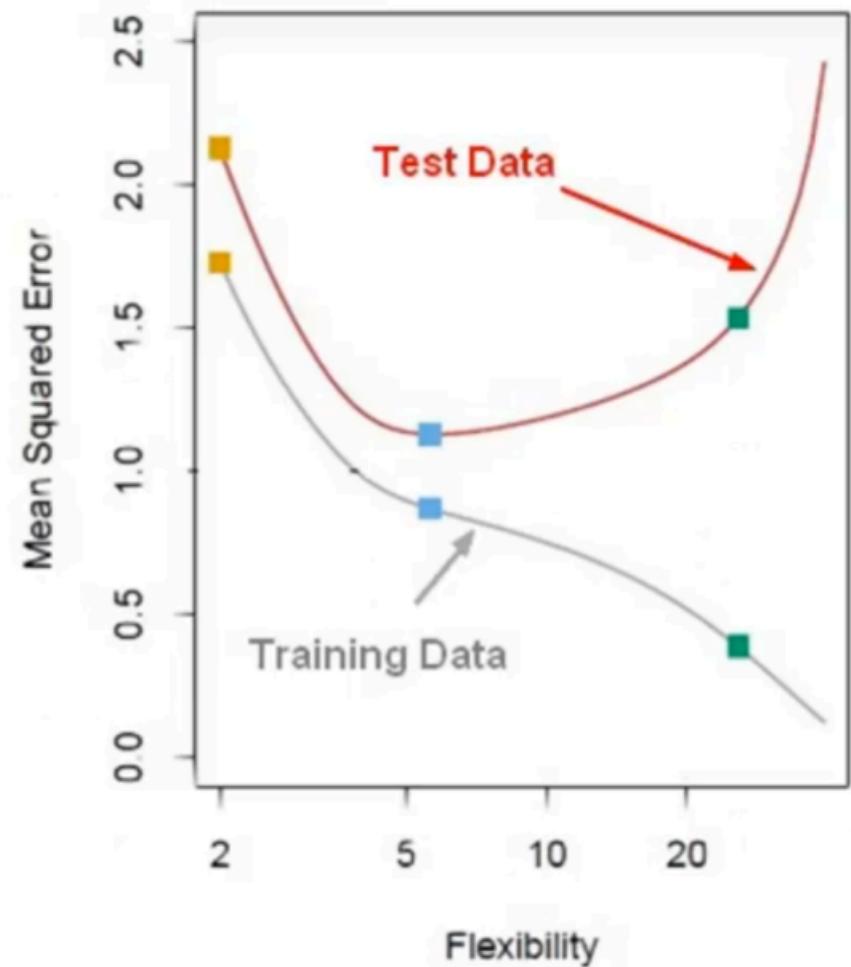
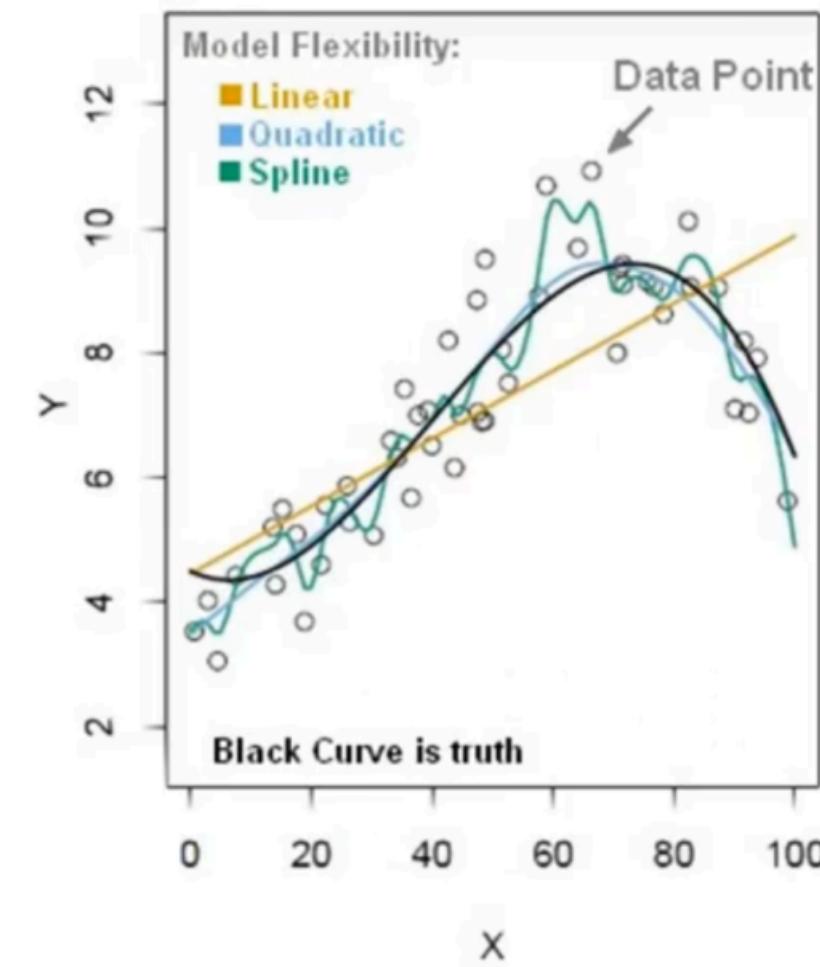
- These different realizations result in a scatter of hits on the target.



- A common temptation for beginners is to continually add complexity to a model until it fits the training set very well.



- Doing this can cause a model to overfit to your training data and cause large errors on new data, such as the test set.
- Let's take a look at an example model on how we can see overfitting occur from a error standpoint using test data!
- We'll use a black curve with some “noise” points off of it to represent the True shape the data follows.



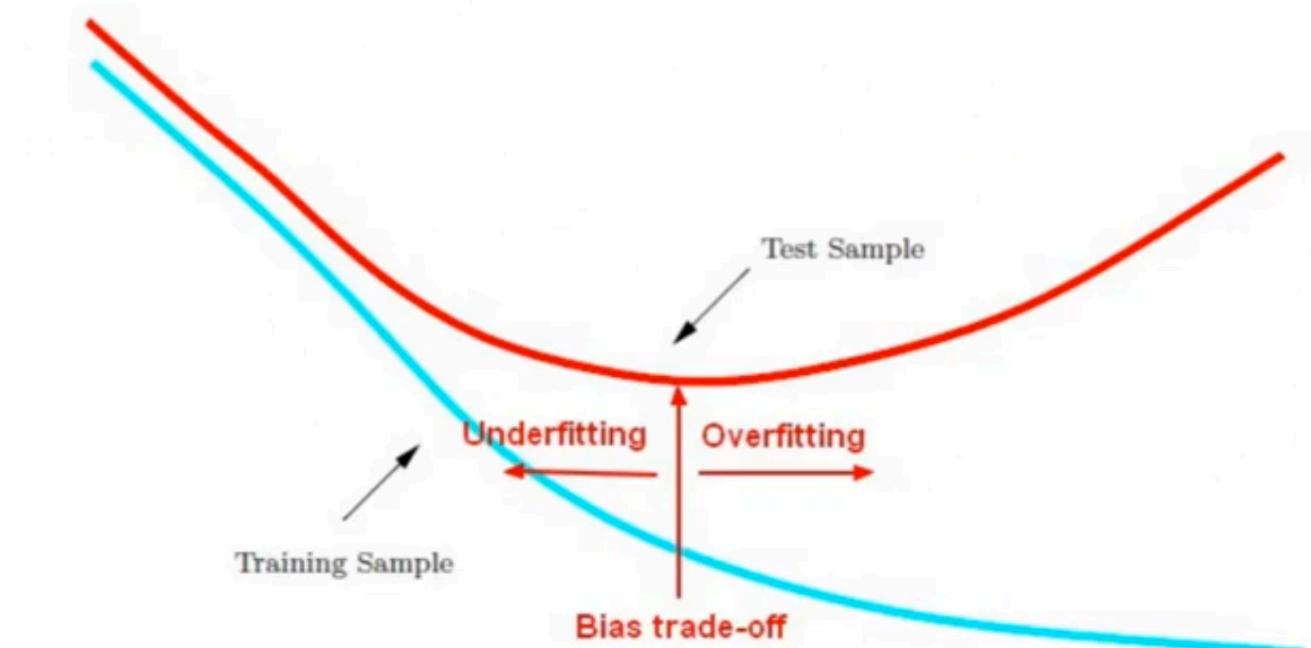
Prediction Error

High Bias

Low Variance

Low Bias

High Variance



Low

High

Model Complexity