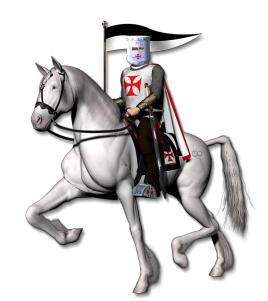


Bias - Variance Tradeoff

Why Bias and Variance are never close together?



What is Bias?





Bias

 Model with high bias pays very little attention to the training data and oversimplifies the model



It's not what I thought

Bias

It always leads to high error on training and test data

You have \$1000 bank balance. Go shopping and treat yourself!



Oops! My bad

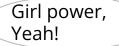
What is Variance?

 The amount by which the model prediction would change if we estimated it using a different training dataset

The results are out! You failed



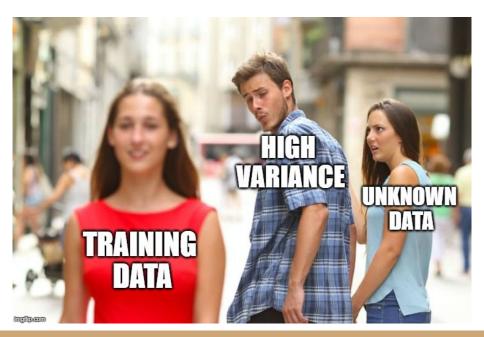
I've tried different
method. You scored
full





Variance

 Model with high variance pays a lot of attention to training data and doesn't generalize on the data which it hasn't seen before



Variance

 As a result, such models perform very well on training data but has high error rates on test data





Noise

- If you were to model the price of an apartment, you know that the price depends on the area of the apartment, no. of bedrooms, etc.
- So those factors contribute to the pattern more bedrooms would typically lead to higher prices.
- However, all apartments with the same area and no. of bedrooms do not have the exact same price.
- The variation in price is the noise.





Noise

- Consider driving.
- Given a curve with a specific curvature, there is an optimal direction of steering and an optimal speed.
- When you observe 100 drivers on that curve, most of them would be close to that optimal steering angle and speed.
- But they will not have the exact same steering angle and speed.
- So again, the curvature of the road contributes to the pattern for steering angle and speed, and then there is noise causing deviations from this optimal value.



Archery Competition



Perfect Model

Bull's eye!



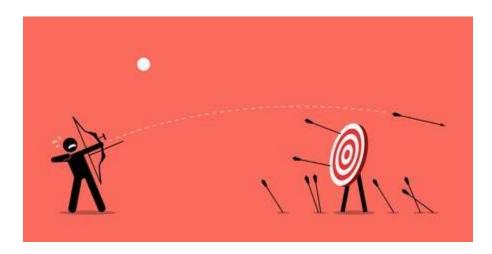
Other possibilities

- Try hitting away from centre:
- That's how much your prediction(model) got worse.



Other Possibilities

- 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 the model

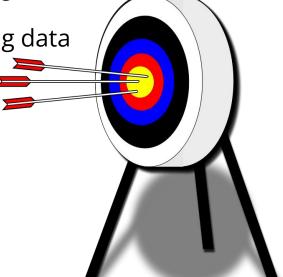


Low Bias & Low Variance

 This would have to be the ideal situation which we would like for our Machine Learning model

The error of prediction should be as less as possible

Predictions shouldn't change with change in training data



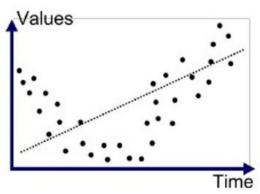
High Bias & High Variance

- This would have to be the worst situation which we would like for our Machine Learning models
- The error of prediction or bias is very high
- The predictions also fluctuate rapidly with change in training data



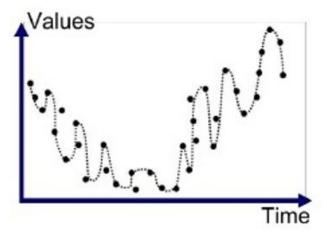
High Bias & Low Variance

- Underfitting happens when model is unable to capture the underlying pattern of data
- Happens when we have very less amount of data to build an accurate model
- Also happens when we try to build a linear model with non-linear data
- Eg.: Linear, Logistic Regression



Low Bias & High Variance

- Overfitting happens when our model captures the noise around the underlying pattern in data
- Happens when we train our model a lot on a noisy dataset
- Happens in complex models like decision trees, SVMs



Overfitting

- The goal of machine learning is to model the pattern and ignore the noise.
- Anytime an algorithm is trying to fit the noise in addition to the pattern, it is overfitting.

Why Bias-Variance Tradeoff

- If our model is too simple and has very few parameters, then it may have high bias and low variance.
- On the other hand if our model has large number of parameters then it's going to have high variance and low bias
- So we need to find the right/good balance without overfitting and underfitting the data.

Managing Bias & Variance

A proper Machine Learning workflow includes:

- Separate training and test data sets
- Trying appropriate algorithms
- Fitting model parameters
- Tuning impactful hyperparameters
- Proper performance metrics
- Systematic cross-validation



Be a good boy

