

Unit 1 Regression Modeling

Variance-Bias Trade off

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Introduction

predictive model: it is important to understand prediction errors (bias and variance)
gaining a proper understanding of these errors would help us not only to build accurate models but also to avoid the mistake of overfitting and underfitting.

start w/ the basics and see how they make difference to models.

What is bias?

the difference btw the average prediction of model and the correct value, which we are trying to predict.

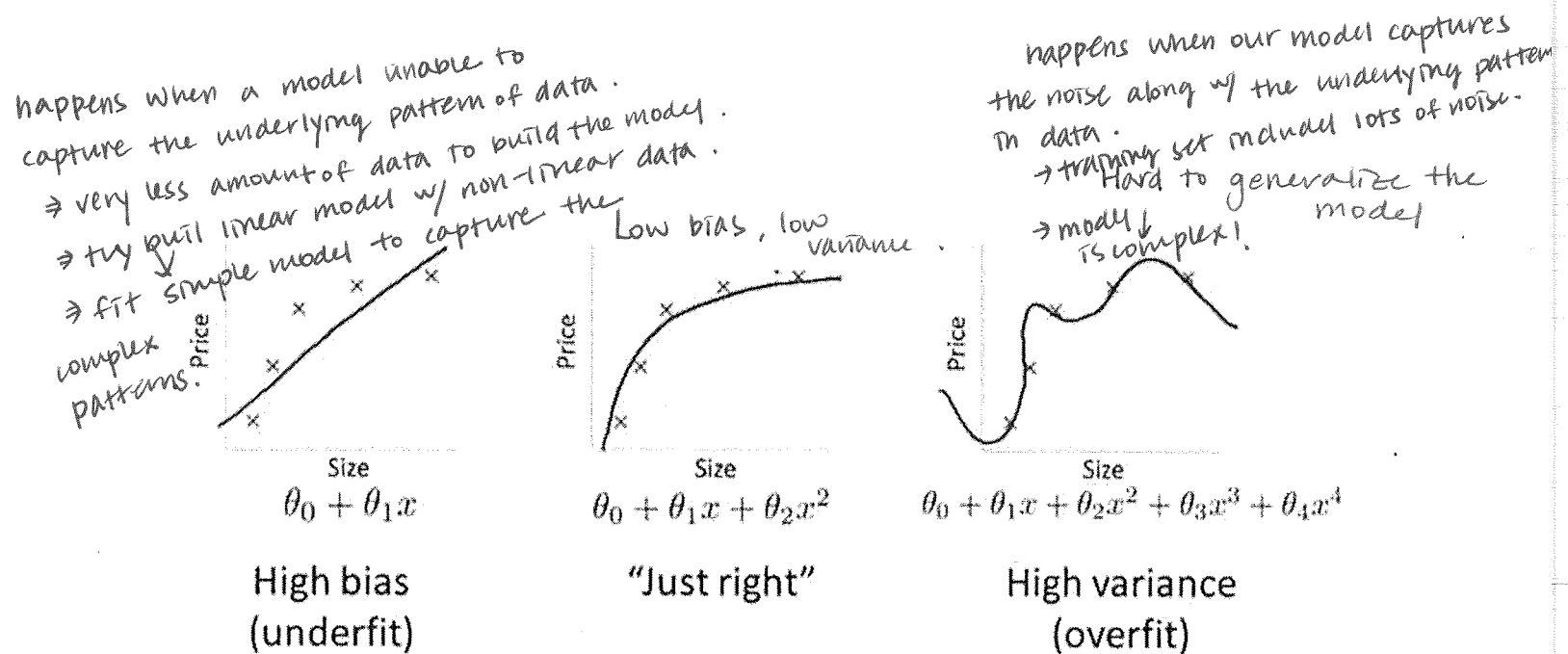
$$E(y - \hat{y})$$

model w/ high bias \rightarrow under fitting.

What is variance?

variability of model prediction for a given data point. (spread of data)

models w/ high variance. perform very well on training data but has high error rates on test data.



Error = Bias² + variance + Irreducible Error.

"data" \uparrow under- or over-fitted model

model selection. \uparrow

no matter how good model we make, cannot be reduced by creating good models (noise of data).

There is a trade off between a model's ability to minimize bias and variance.

Why is Bias Variance Trade off?

too simple model (has very few parameters) \Rightarrow High bias and low variance.

larger # of parameters \Rightarrow High variance and low bias.

\Rightarrow Need to find the right/good balance w/o overfitting and underfitting the data.

This tradeoff in complexity is why there is a tradeoff btw bias and variance.

$$\text{Total Error} = \text{Bias}^2 + \text{variance} + \text{Irreducible error}$$

\uparrow \downarrow fixed.

symptoms

High variance. ① Training error is much lower than test error.

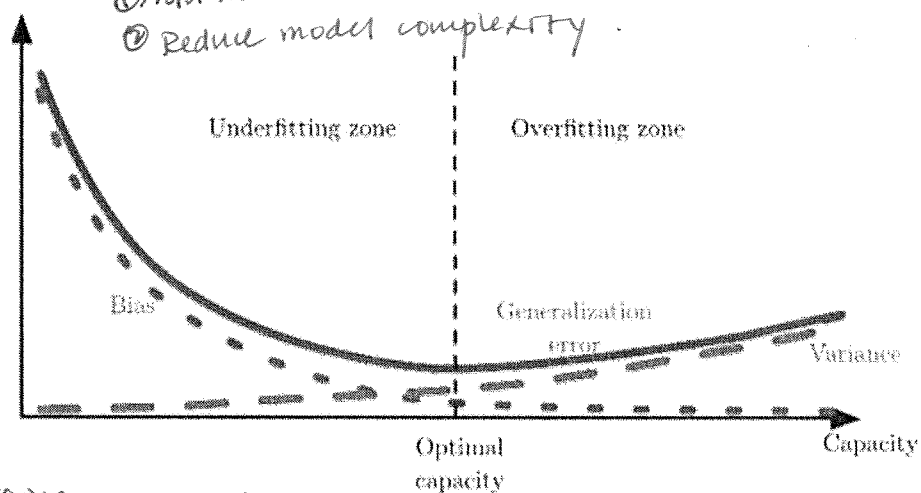
② Training error is lower than ϵ .

③ test error is above ϵ .

solution \downarrow

① Add more training data

② Reduce model complexity.



High Bias : Training error $> \epsilon$.

① complex model (ex) non-linear₂ model).

② Add features.

③ Boosting.