Bias - Variance and Error.

1. Bias - Voriance Tradeoff.

Mean squared error: $\mathbb{E}_{(x,y)} \sim \text{test set } |\hat{f}(x) - y|^2$ too high.

- · Overfitting: the model is to clusely related to the training set, thu not so generalize to other examples.
- · Underfitting: the model doesn't gother enough infromation from training set thus con't predict right result. (doesn't capture the link between features and targets)
- · Data is simply noisy, too much noise in the dataset

1.2.
$$MSE = \mathbb{E} \left(\frac{y - \hat{f}(x)^2}{x^2} \right)$$

= $\mathbb{E} \left(\frac{(E + f(x) - \hat{f}(x)^2}{x^2} \right)$
= $\mathbb{E} \left(\frac{(E + f(x) - \hat{f}(x)^2}{x^2} \right) + V_{av} \left(\frac{f(x) - f(x)}{x^2} \right)$

$$= \sigma^2 + (\beta i \alpha s \hat{f}(x))^2 + V_{ar}(\hat{f}(x))$$

High Bias: Under fitting
High Vor: Over fitting

2. Error analysis & ablative analysis.