## Deciding What to Do Next Revisited

Our decision process can be broken down as follows:

- **Getting more training examples:** Fixes high variance
- Trying smaller sets of features: Fixes high variance
- Adding features: Fixes high bias
- Adding polynomial features: Fixes high bias
- **Decreasing**  $\lambda$ : Fixes high bias
- Increasing λ: Fixes high variance.

## **Diagnosing Neural Networks**

- A neural network with fewer parameters is **prone to underfitting.** It is also **computationally cheaper.**
- A large neural network with more parameters is **prone to overfitting**. It is also **computationally expensive.** In this case you can use regularization (increase  $\lambda$ ) to address the overfitting.

Using a single hidden layer is a good starting default. You can train your neural network on a number of hidden layers using your cross validation set. You can then select the one that performs best.

## **Model Complexity Effects:**

- Lower-order polynomials (low model complexity) have high bias and low variance. In this case, the model fits poorly consistently.
- Higher-order polynomials (high model complexity) fit the training data extremely well
  and the test data extremely poorly. These have low bias on the training data, but very
  high variance.
- In reality, we would want to choose a model somewhere in between, that can generalize well but also fits the data reasonably well.