

Machine Learning Course - CS-433

# Overfitting

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minor changes by Martin Jaggi 2016



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# Motivation

Most models can be either *too limited*, or also *too powerful*. In other words, models can either underfit or overfit.

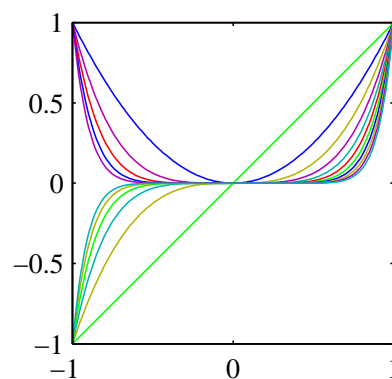
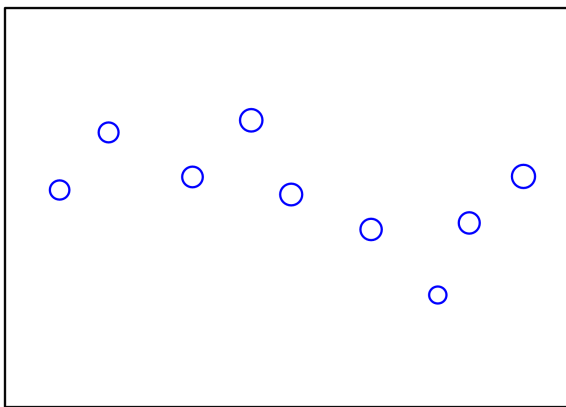
## Can Linear Models Overfit?

**Yes! #1:** Consider simple linear regression. Given one-dimensional input  $x_n$ , we can generate a polynomial basis.

$$\phi(x_n) = [1, x_n, x_n^2, x_n^3, \dots, x_n^M]$$

Then we fit a linear model on the generated features, instead of the original feature:

$$y_n \approx w_0 + w_1 x_n + w_2 x_n^2 + \dots + w_M x_n^M$$



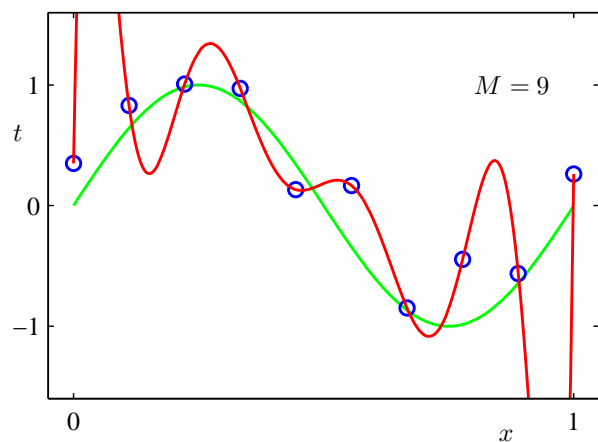
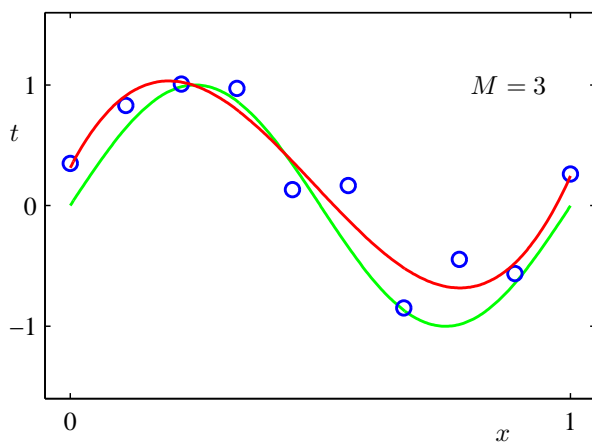
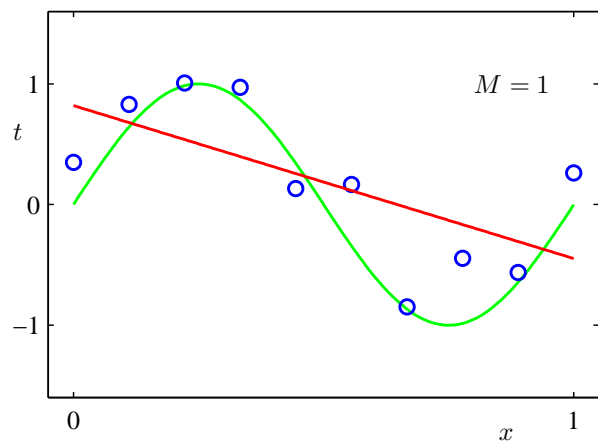
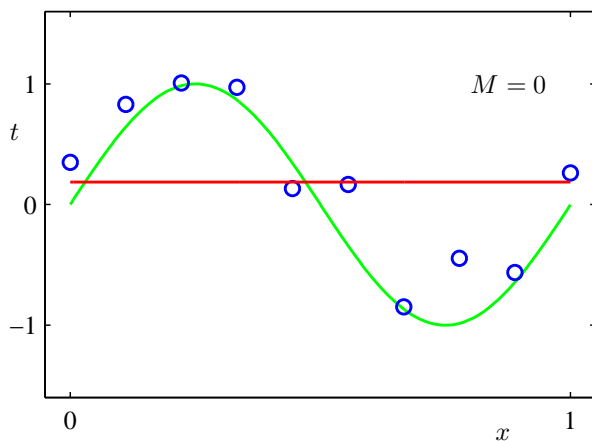
# Overfitting and Underfitting

**Overfitting** is fitting the noise in addition to the signal. **Underfitting** is not fitting the signal well.

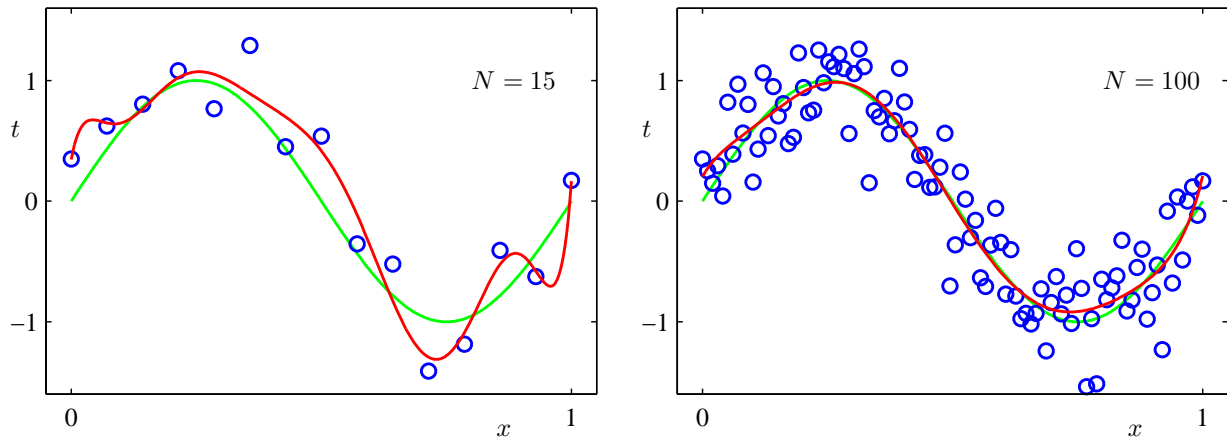
In real life, it is difficult to distinguish signal from noise.

## Complex Models Overfit Easily

Circles are data points, green line is the truth & red line is the model fit.  $M$  is the maximum degree in the generated polynomial basis.



If you increase the amount of data (increase  $N$ , but keep  $M$  fixed), overfitting *might* reduce.



## Preventing Overfitting

- Get more data!
- Regularization - Force the model to be not too complex

# Occam's Razor

One solution is dictated by [Occam's razor](#) which states that “Plurality is not to be posited without necessity” or rephrased “Simpler models are better - only use complicated ones if strictly necessary”.

Sometimes, if you increase the amount of data, you might reduce overfitting. But, when unsure, choose a simple model over a complicated one.

We can choose simpler models by adding a [regularization term](#) which ‘penalizes’ complex models.

$$\min_{\mathbf{w}} \quad \frac{1}{2N} \sum_{n=1}^N [y_n - \phi(\mathbf{x}_n)^\top \mathbf{w}]^2 + \Omega(\mathbf{w})$$

where  $\lambda > 0$ .

## ToDo

Read about overfitting in the paper by Pedro Domingos (Sections 3 and 5 of “A few useful things to know about machine learning”).