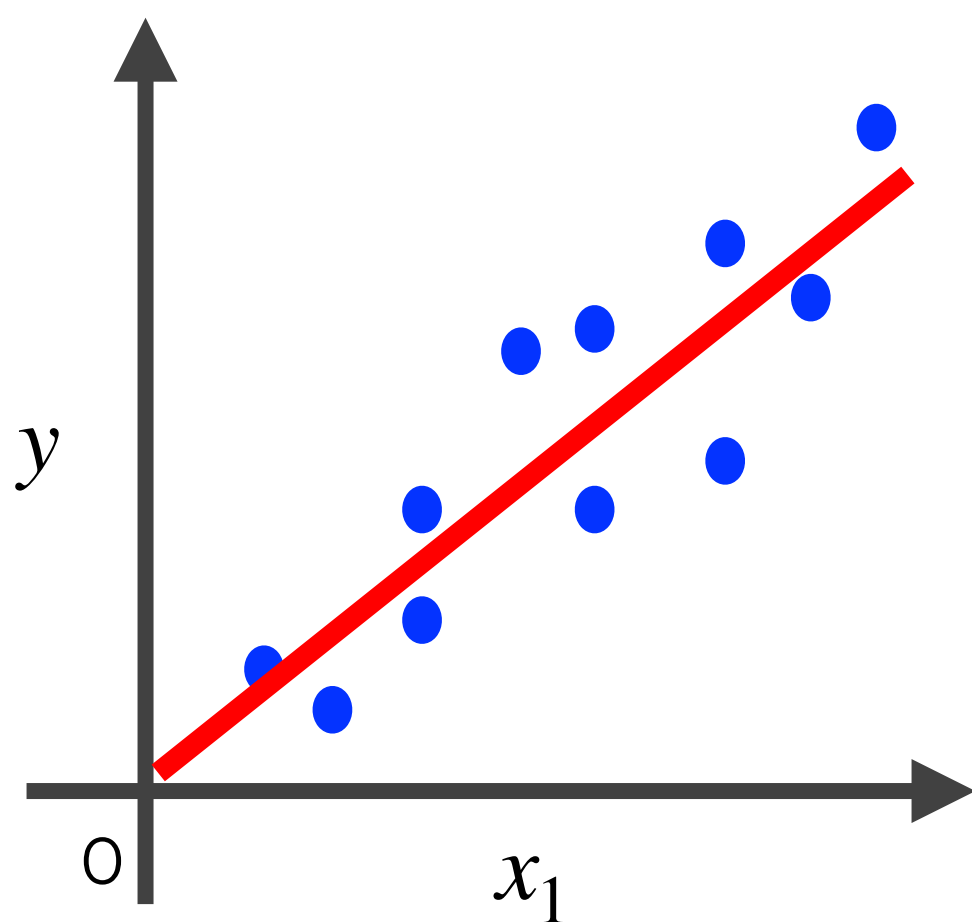


Polynomial Regression, Overfitting and Regularization

Itthi Chatnuntaweck

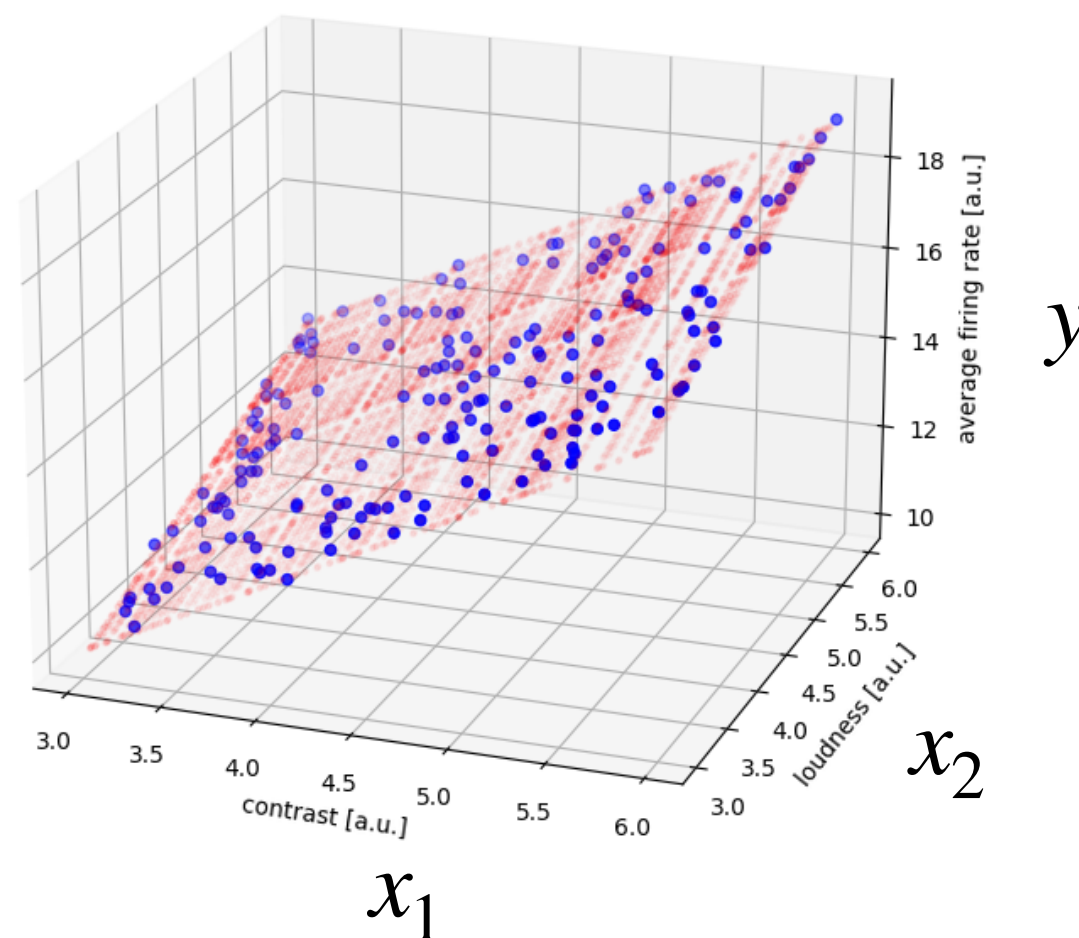
Linear Models

$$y = w_0 + w_1 x_1$$



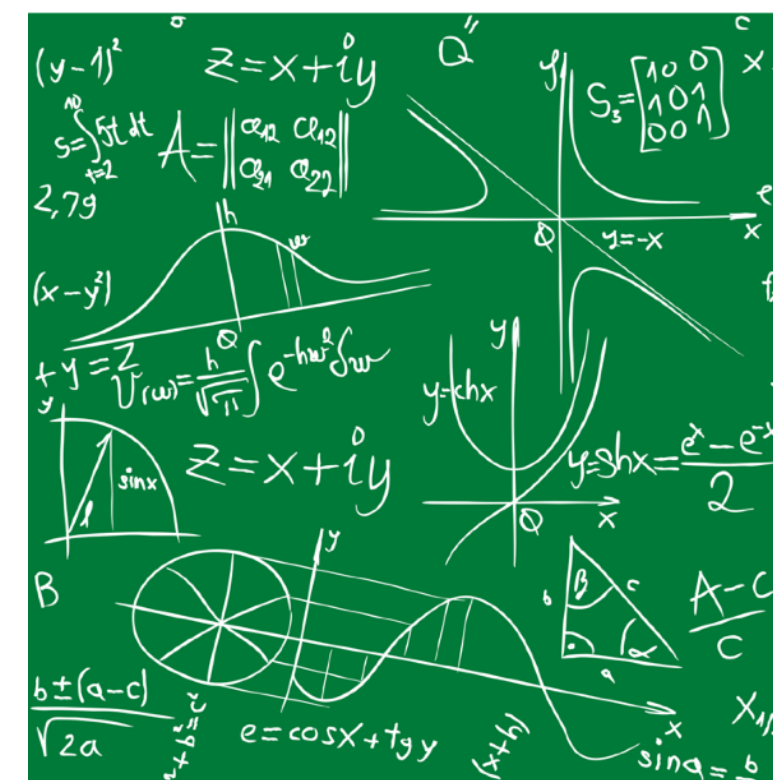
Line
1 feature

$$y = w_0 + w_1 x_1 + w_2 x_2$$



Plane
2 features

$$y = w_0 + w_1 x_1 + w_2 x_2 + \dots + w_p x_p$$

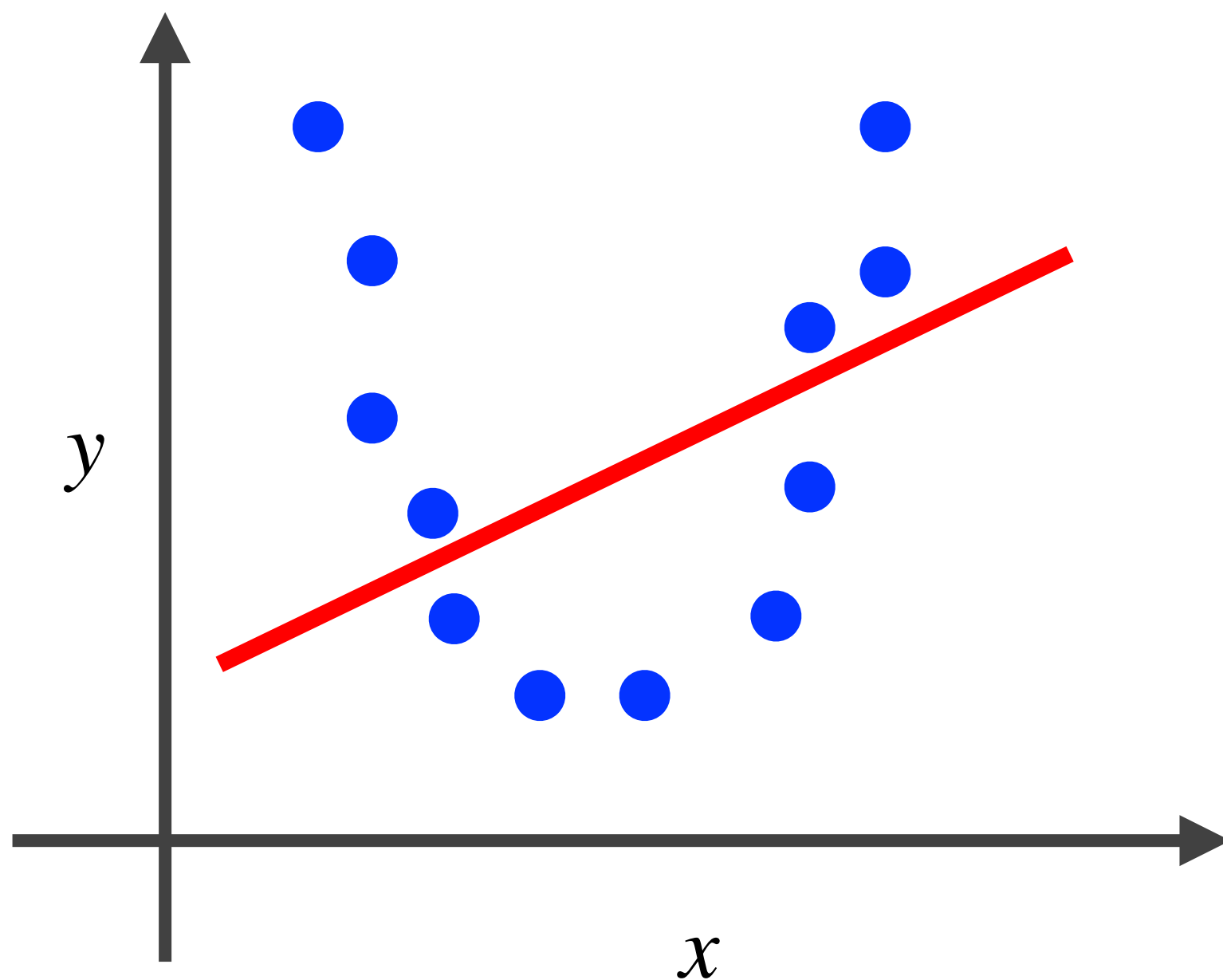


Hyperplane
p features

Nonlinear Relationship

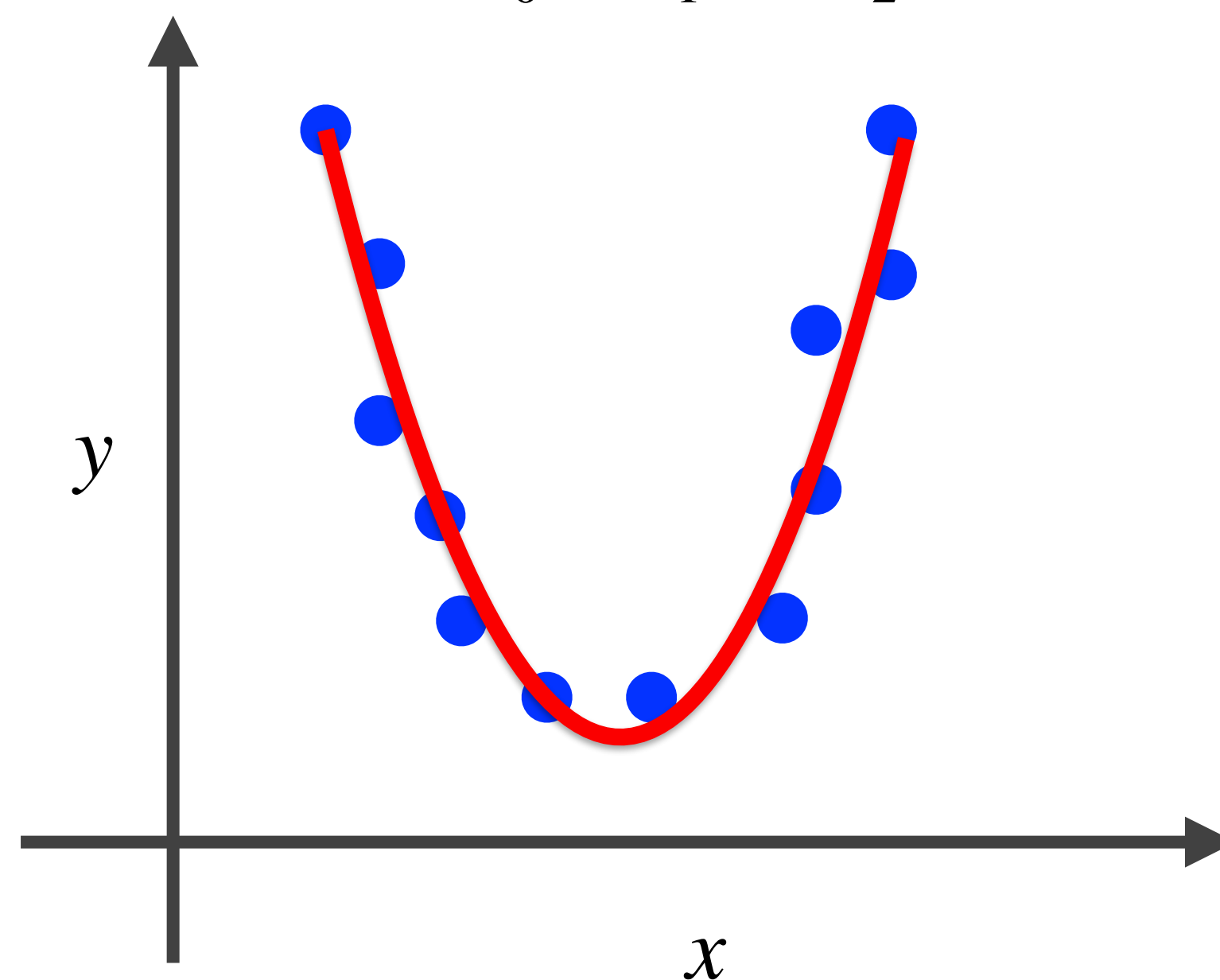
$$y = mx + c$$

$$y = w_0 + w_1x_1$$



$$y = ax^2 + bx + c$$

$$y = w_0 + w_1x + w_2x^2$$



Simple Polynomial

A linear model

$$y = w_0 + w_1x_1 + w_2x_2 + \dots + w_px_p$$

It is linear in both parameters (w_1, w_2, \dots, w_p) and variables (x_1, x_2, \dots, x_p)

There is no nonlinear terms such as w_i^p , w_iw_j , $w_i^kw_j^l$, e^{w_i} and $\log(w_i)$ and x_i^p , x_ix_j , $x_i^kx_j^l$, e^{x_i} and $\log(x_i)$.

A simple polynomial

$$y = w_0 + w_1x + w_2x^2 + \dots + w_px^p$$

It is linear in parameters (w_1, w_2, \dots, w_p) , but not linear in variables because it contains x^2, x^3, \dots, x^p

Simple Polynomial

A linear model

$$\hat{y} = \hat{w}_0 + \hat{w}_1 x_1 + \hat{w}_2 x_2 + \dots + \hat{w}_p x_p$$

A simple polynomial

$$\hat{y} = \hat{w}_0 + \hat{w}_1 x + \hat{w}_2 x^2 + \dots + \hat{w}_p x^p$$

Simple Polynomial

A linear model

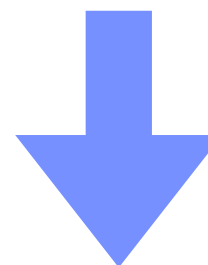
$$\hat{y} = \hat{w}_0 + \hat{w}_1 x_1 + \hat{w}_2 x_2 + \dots + \hat{w}_p x_p$$

feature 1 feature 2 feature p

A simple polynomial

$$\hat{y} = \hat{w}_0 + \hat{w}_1 x + \hat{w}_2 x^2 + \dots + \hat{w}_p x^p$$

Let $z_i = x^i$



$$\hat{y} = \hat{w}_0 + \hat{w}_1 z_1 + \hat{w}_2 z_2 + \dots + \hat{w}_p z_p$$

Simple Polynomial

A linear model

$$\hat{y} = \hat{w}_0 + \hat{w}_1 x_1 + \hat{w}_2 x_2 + \dots + \hat{w}_p x_p$$

feature 1 feature 2 feature p

A simple polynomial

$$\hat{y} = \hat{w}_0 + \hat{w}_1 z_1 + \hat{w}_2 z_2 + \dots + \hat{w}_p z_p$$

sklearn.preprocessing.PolynomialFeatures

```
class sklearn.preprocessing.PolynomialFeatures(degree=2, *, interaction_only=False, include_bias=True, order='C')
```

[\[source\]](#)

sklearn.linear_model.LinearRegression

```
class sklearn.linear_model.LinearRegression(*, fit_intercept=True, copy_X=True, n_jobs=None, positive=False)
```

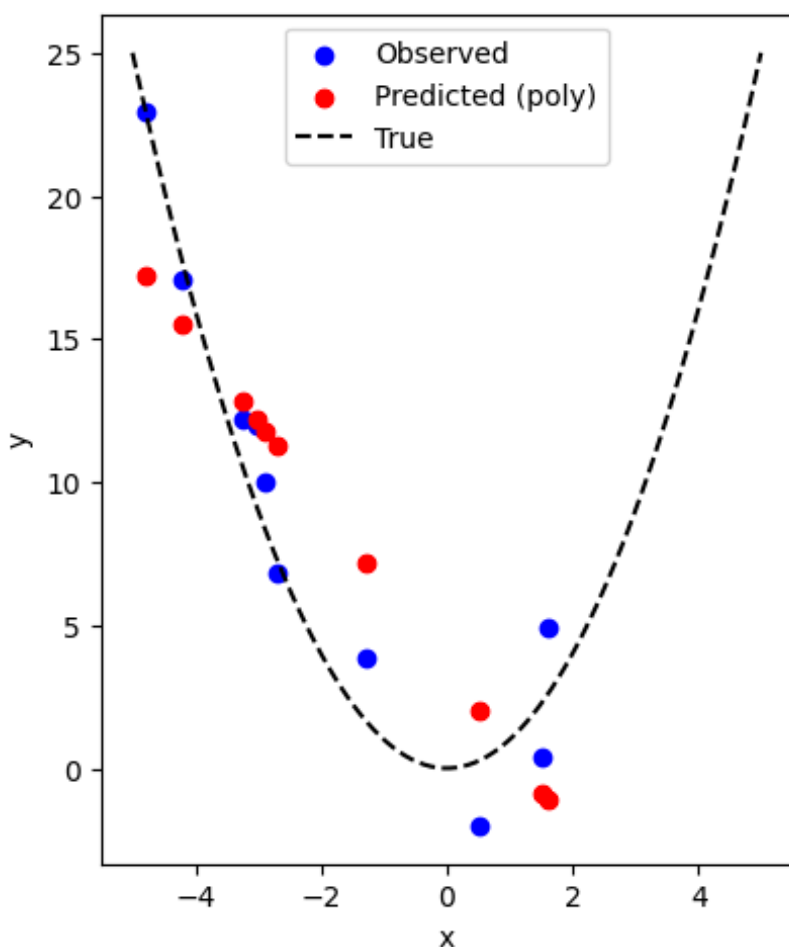
[\[source\]](#)

Underfitting and Overfitting

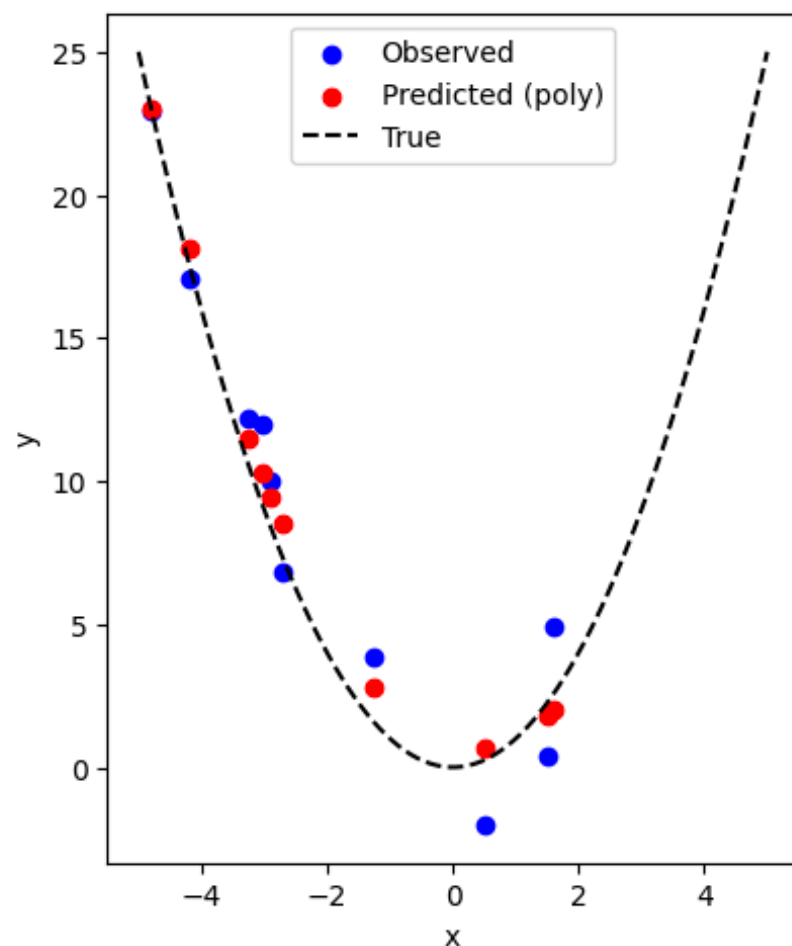
Assume that the true relationship is $y = x^2$, and we have collected only 10 points

$$\hat{y} = \hat{w}_0 + \hat{w}_1x + \hat{w}_2x^2 + \dots + \hat{w}_px^p$$

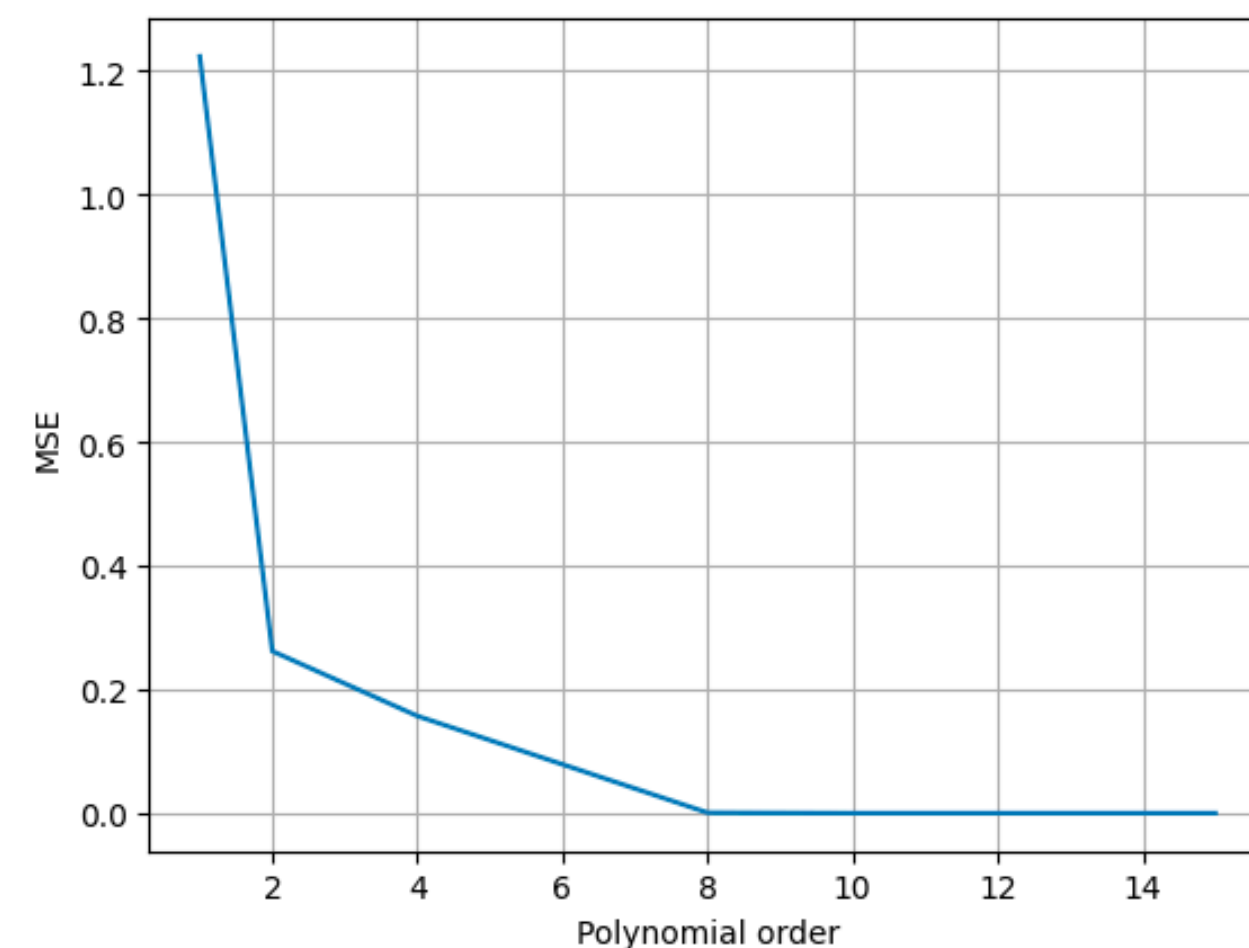
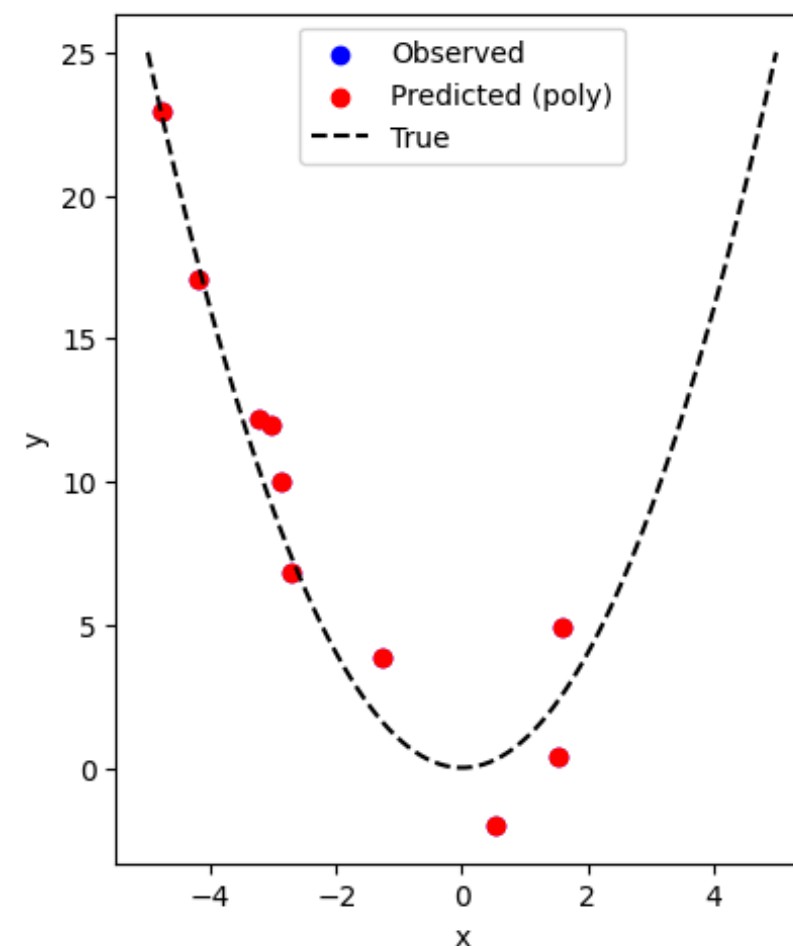
Order = 1 with MSE = 1.2237



Order = 2 with MSE = 0.2620



Order = 15 with MSE = 0.0000



Underfitting and Overfitting

Assume that the true relationship is $y = x^2$, and we have collected only 10 points

$$\hat{y} = \hat{w}_0 + \hat{w}_1x + \hat{w}_2x^2 + \dots + \hat{w}_px^p$$

