## Class 7- Polynomial Regression



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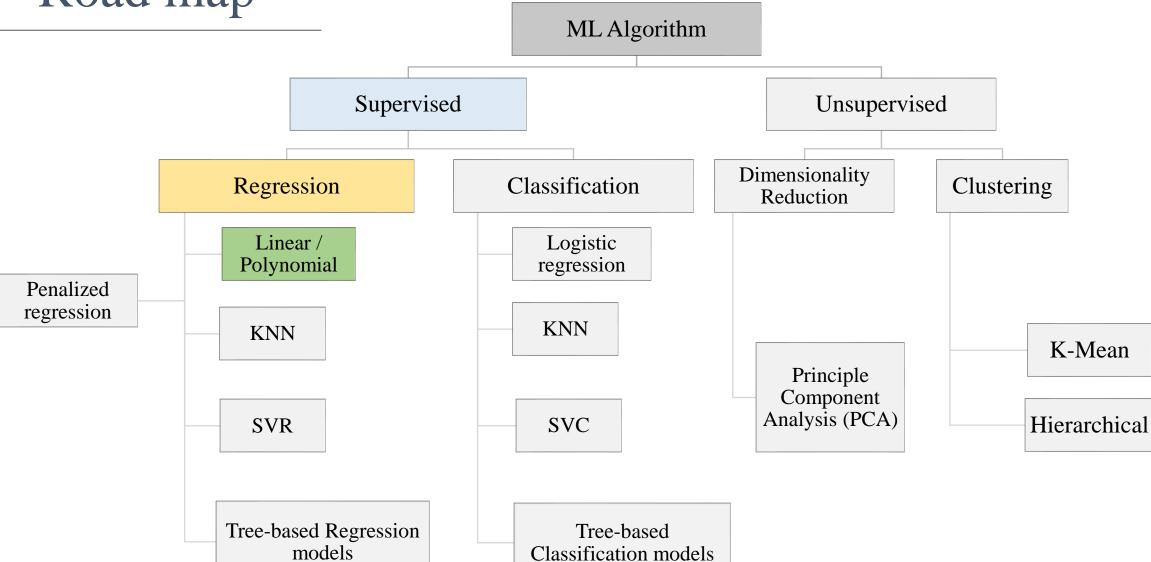








# Road map





13



# Polynomial regression model

The polynomial regression model is a <u>special case</u> of multiple linear regression models!

- Create new variables  $X_1 = X$ ,  $X_2 = X^2$ , ... etc and then treat as multiple linear regression.
- Not really interested in the coefficients; more interested in the fitted function!

$$\hat{f}(X) = f_{w,b}(X) = b + w_1 x + w_2 x^2 + \dots + w_d x^d$$

- W is a d-dimensional vector of parameters
- b is a real number
- d is the polynomial degree of the model (we either fix the d at some reasonably low value, else use **cross-validation** to choose d



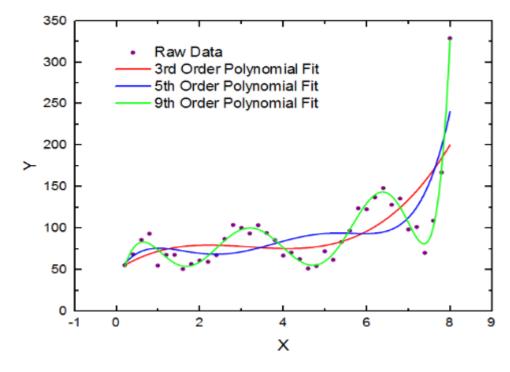


## The optimization problem

The optimization problem is defined as:

$$Min_{w,b} MSE = \frac{1}{N} \sum_{i=1}^{N} (y_i - \widehat{y}_i)^2 = \frac{1}{N} \sum_{i=1}^{N} (y_i - f_{w,b}(X_i))^2$$

- We use the same **loss function** as in linear regression!
- The solution to this optimization problem is  $w^*$  and  $b^*$
- Now we can make predictions!





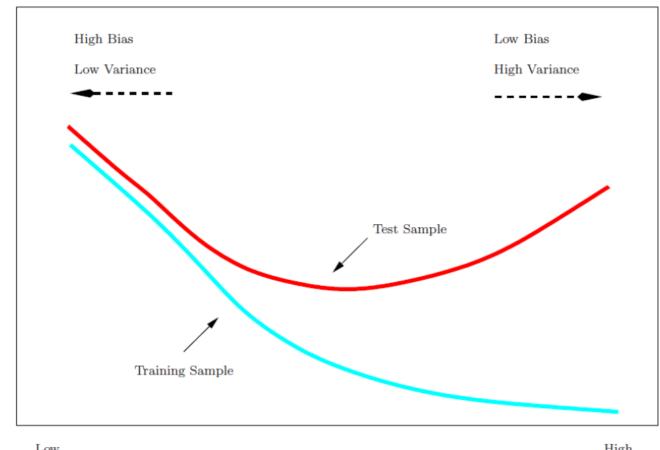


#### Polynomial Regression Evaluation Metrics

$$MSE = \frac{1}{N} \sum_{i=1}^{N} (y_i - \hat{y})^2$$

$$RMSE = \sqrt{MSE}$$

Prediction Error



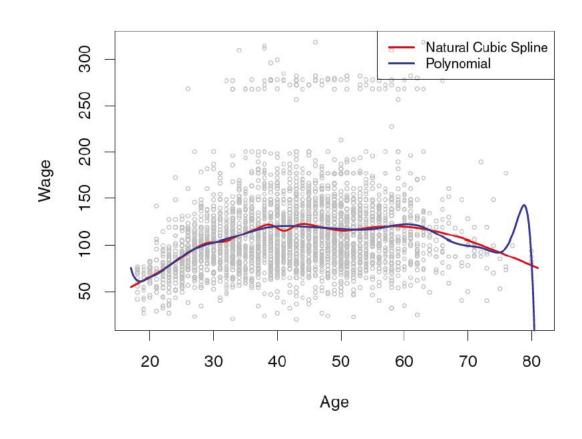
Low High

Model Complexity



### Caveats!

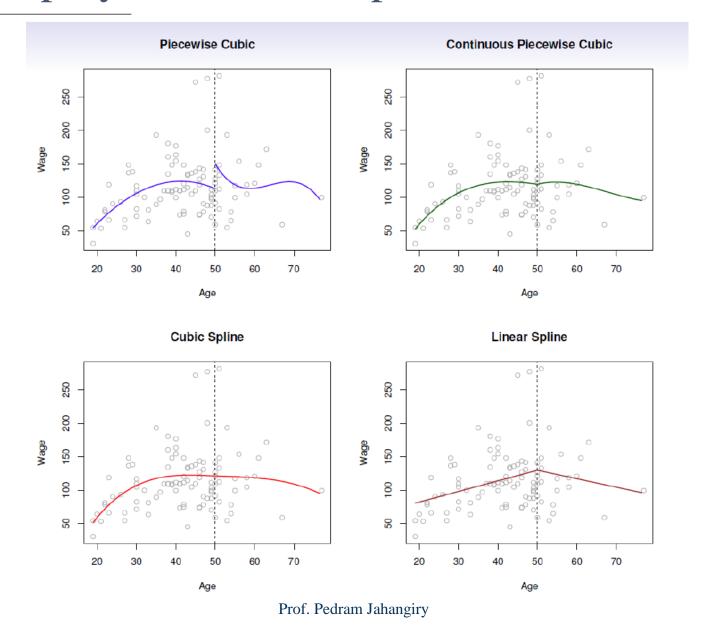
- In general quadratic loss functions are sensitive to outliers
- Polynomials have notorious tail behavior
- Polynomials are global fit!
- Solution: piecewise polynomial, splines and local regressions.







## Piecewise polynomials and splines!





# Students' questions

1) TBA







