

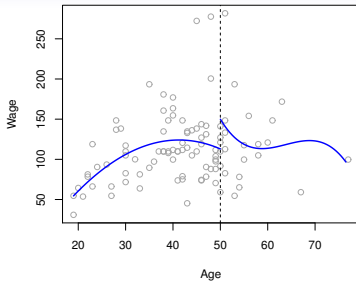
# Piecewise Polynomials

- Instead of a single polynomial in  $X$  over its whole domain, we can rather use different polynomials in regions defined by knots. E.g. (see figure)

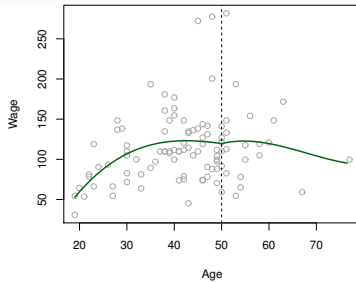
$$y_i = \begin{cases} \beta_{01} + \beta_{11}x_i + \beta_{21}x_i^2 + \beta_{31}x_i^3 + \epsilon_i & \text{if } x_i < c; \\ \beta_{02} + \beta_{12}x_i + \beta_{22}x_i^2 + \beta_{32}x_i^3 + \epsilon_i & \text{if } x_i \geq c. \end{cases}$$

- Better to add constraints to the polynomials, e.g. continuity.
- *Splines* have the “maximum” amount of continuity.

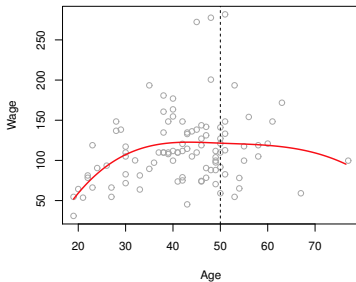
**Piecewise Cubic**



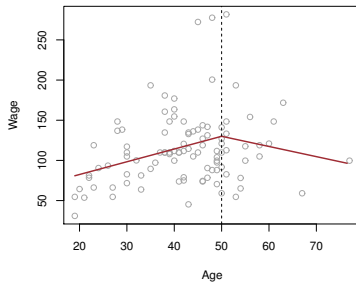
**Continuous Piecewise Cubic**



**Cubic Spline**



**Linear Spline**



## Linear Splines

*A linear spline with knots at  $\xi_k$ ,  $k = 1, \dots, K$  is a piecewise linear polynomial continuous at each knot.*

We can represent this model as

$$y_i = \beta_0 + \beta_1 b_1(x_i) + \beta_2 b_2(x_i) + \cdots + \beta_{K+1} b_{K+1}(x_i) + \epsilon_i,$$

where the  $b_k$  are *basis functions*.

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$$\begin{aligned} b_1(x_i) &= x_i \\ b_{k+1}(x_i) &= (x_i - \xi_k)_+, \quad k = 1, \dots, K \end{aligned}$$

Here the  $()_+$  means *positive part*, i.e.

$$(x_i - \xi_k)_+ = \begin{cases} x_i - \xi_k & \text{if } x_i > \xi_k \\ 0 & \text{otherwise} \end{cases}$$

