

# Ch 7.4: Cubic splines

## Lecture 24 - CMSE 381

Prof. Elizabeth Munch

Michigan State University

::

Dept of Computational Mathematics, Science & Engineering

Weds, Nov 6, 2023

## **Last time:**

- 7.2 Step functions
- 7.3 Basis functions

## **This lecture:**

- 7.4 Cubic splines

## **Announcements:**

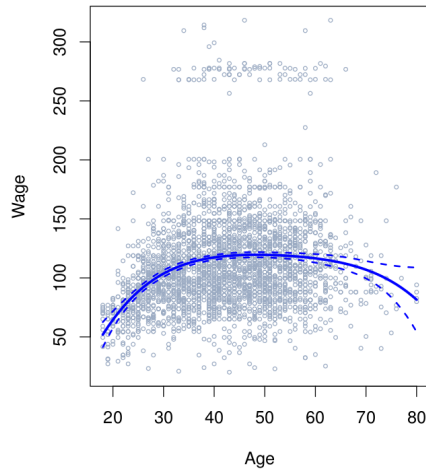
- Homework # 6 is now due Wednesday

# Section 1

Last time

# Polynomial regression

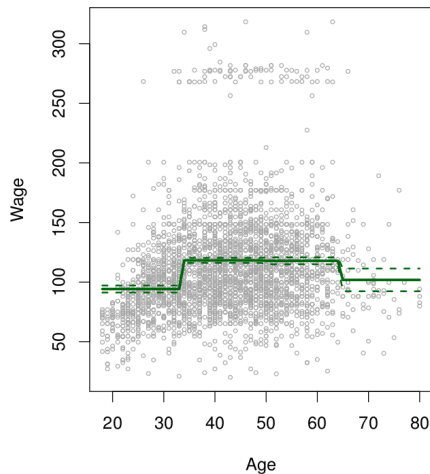
$$y_i = \beta_0 + \beta_1 x_1 + \beta_2 x_i^2 + \cdots + \beta_d x_i^d + \varepsilon_i$$



# Step function regression

$$\begin{aligned}C_0(X) &= I(X < c_1), \\C_1(X) &= I(c_1 \leq X < c_2), \\C_2(X) &= I(c_2 \leq X < c_3), \\&\vdots \\C_{K-1}(X) &= I(c_{K-1} \leq X < c_K), \\C_K(X) &= I(c_K \leq X),\end{aligned}$$

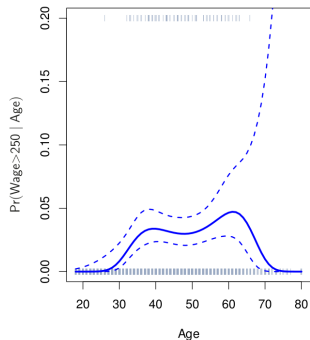
$$y_i = \beta_0 + \beta_1 C_1(x_i) + \beta_2 C_2(x_i) + \cdots + \beta_K C_K(x_i) + \varepsilon_i$$



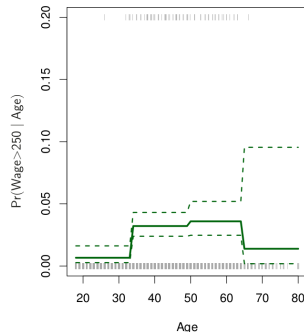
# Classification version

$$\Pr(y_i > 250 \mid x_i) =$$

$$\frac{\exp(\beta_0 + \beta_1 x_i + \cdots + \beta_d x_i^d)}{1 + \exp(\beta_0 + \beta_1 x_i + \cdots + \beta_d x_i^d)}$$



$$\frac{\exp(\beta_0 + \beta_1 C_1(x_i) + \beta_2 C_2(x_i) + \cdots + \beta_K C_K(x_i))}{1 + \exp(\beta_0 + \beta_1 C_1(x_i) + \beta_2 C_2(x_i) + \cdots + \beta_K C_K(x_i))}$$



# Basis Functions Setup

Polynomial and piecewise-constant regression models are special cases of a *basis function* approach.

$$y_i = \beta_0 + \beta_1 b_1(x_i) + \beta_2 b_2(x_i) + \cdots + \beta_K b_K(x_i) + \varepsilon_i$$

## Section 2

### Regression Splines



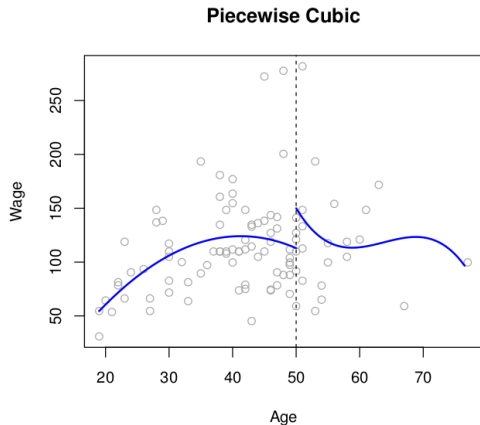
# Piecewise polynomials

- Fit a polynomial regression

$$y_i = \beta_0 + \beta_1 x_i + \beta_2 x_i^2 + \cdots + \beta_d x_i^d + \varepsilon_i$$

- Let the  $\beta_i$ 's be different at different locations of the range.

# Example of piecewise polynomial

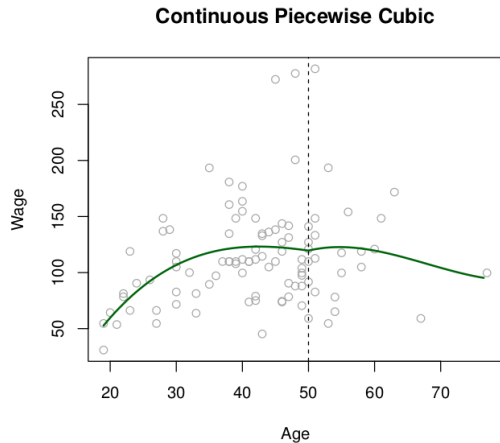


Example:

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

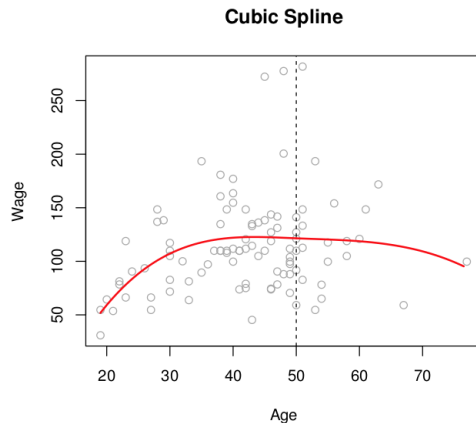
# The fix

- Fit piecewise polynomial
- Require continuity at knots



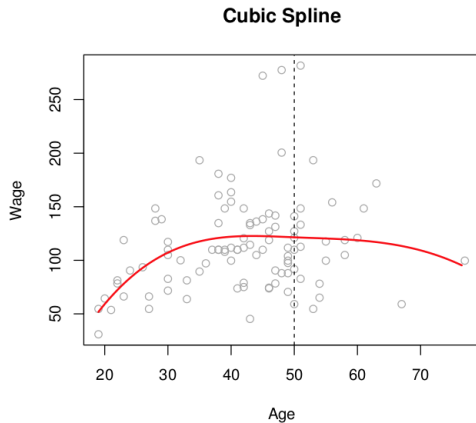
# The better fix: Cubic splines

- Fit piecewise polynomial
- Require continuity at knots
- Require the first and second derivatives to be continuous at knots



# Cubic splines: degrees of freedom

$$f(x) = \begin{cases} \beta_0^1 + \beta_1^1 x + \beta_2^1 x^2 + \beta_3^1 x^3 & x < c \\ \beta_0^2 + \beta_1^2 x + \beta_2^2 x^2 + \beta_3^2 x^3 & x > c \end{cases}$$



# Spline basis representation

Want to pick  $b_i$  so that we represent a cubic spline with  $K$  knots as

$$y_i = \beta_0 + \beta_1 b_1(x_i) + \beta_2 b_2(x_i) + \cdots + \beta_{K+3} b_{K+3}(x_i) + \varepsilon_i$$

# Truncated power basis function

$$h(x, z) = (x - z)_+^3 = \begin{cases} (x - z)^3 & \text{if } x > z \\ 0 & \text{else} \end{cases}$$

Desmos link: <https://www.desmos.com/calculator/esucuulbgj>

# The basis for cubic splines

Given knots at  $z_1, \dots, z_K$

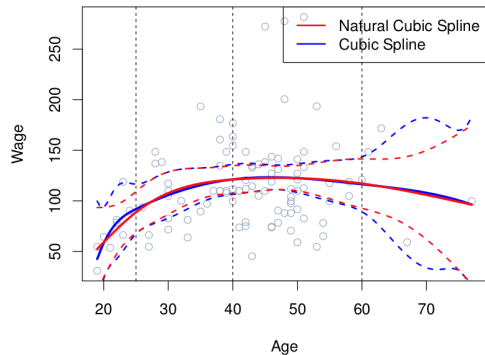
- $X$
- $X^2$
- $X^3$
- $h(X, z_1)$
- $h(X, z_2)$
- $\vdots$
- $h(X, z_K)$

$$f(X) = \beta_0 + \beta_1 X + \beta_2 X^2 + \beta_3 X^3 + \beta_4 h(X, z_1) + \beta_5 h(X, z_2) + \dots + \beta_{k+3} h(X, z_K)$$

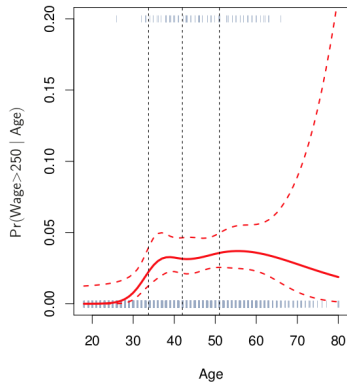
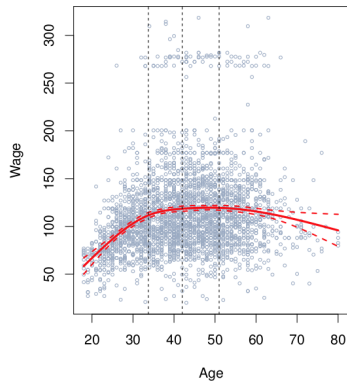


# Coding example

# Notes on cubic splines

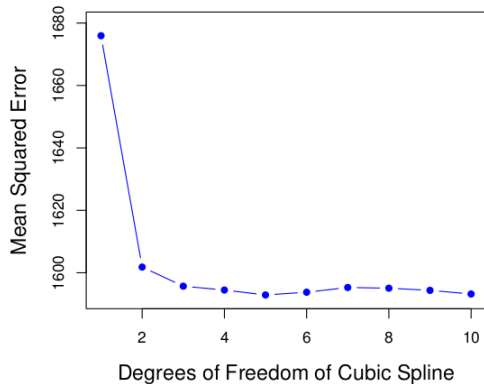


# Where to put the knots?

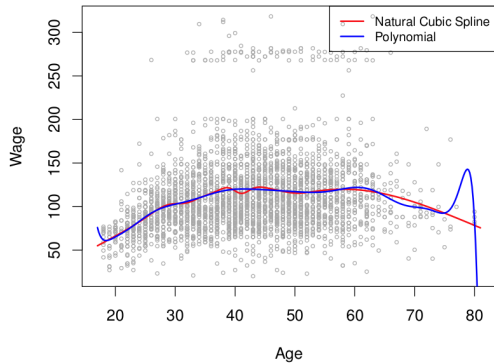


# How many knots to use?

When in doubt, Cross-Validate.



# Cubic splines vs Polynomial Regression



# Next time

Status	Lec #	Date		Reading	Homeworks
		Mon	Oct 23	No class - Fall break	
		Wed	Oct 25	<b>Midterm #2</b>	
Done	20	Fri	Oct 27	Dimension Reduction	6.3
Done	21	Mon	Oct 30	More dimension reduction; High dimensions	6.4
Done	22	Wed	Nov 1	Polynomial & Step Functions	7.1, 7.2
Pushed	23	Fri	Nov 3	Step Functions; Basis functions; Start Splines	7.2 - 7.4
	24	Mon	Nov 6	Regression Splines	7.4
	25	Wed	Nov 8	Decision Trees	8.1
	26	Fri	Nov 10	Random Forests	8.2.1, 8.2.2
	27	Mon	Nov 13	Maximal Margin Classifier	9.1
	28	Wed	Nov 15	SVC	9.2
	29	Fri	Nov 17	SVM	9.3, 9.4
	30	Mon	Nov 20	Single layer NN	10.1
	31	Wed	Nov 22	Virtual: Project office hours	
		Fri	Nov 24	No class - Thanksgiving	
		Mon	Nov 27	<b>Review</b>	
		Wed	Nov 29	<b>Midterm #3</b>	
	32	Fri	Dec 1	Multi Layer NN	10.2
	33	Mon	Dec 4	CNN	10.3
	34	Wed	Dec 6	Unsupervised Learning & Clustering	12.1, 12.4
	35	Fri	Dec 8	Virtual: Project office hours	Project due