Ch. 7 Nonlinear Regression - Problem Bank Questions

October 24, 2020

- 1. Consider the step function model in equation (7.5) of ISLR. State whether each of the below is true or false, and explain your answer.
- (a) T/F: In step function regression, β_0 can be interpreted as the mean value for Y when $X < c_1$.
- (b) T/F: $\mathbf{1}_{X_i>0} = 1$ when $X_i = -10$.
- (c) T/F: Step function regression always has lower bias than linear regression.
- (d) T/F: β_j represents the average value of X_j .

- 2. Regression splines: For the true and false, explain your answer.
- (a) T/F: As the number of knots in polynomial regression increases, the variance decreases.
- (b) T/F: Polynomial regression splines can be discontinuous without constraints.
- (c) T/F: Adding constraints to the piecewise polynomials that are fit increases the flexibility of the model.
- (d) T/F: A cubic spline with K knots requires 3 + K degrees of freedom.
- (e) T/F: Splines are most flexible in regions with many knots.
- (f) T/F: Splines and polynomial regression models always fit well at the boundaries of the X values.

- 3. Smoothing Splines and Local Regression: For the true and false, explain your answer.
- (a) T/F: The degrees of freedom refers to the number of coefficients in a model.
- (b) T/F: The degrees of freedom is a measure of the flexibility of a model.
- (c) T/F: The higher the degrees of freedom, the higher the bias of the model.
- (d) T/F: The smoothing spline will have knots at the location of each of the training points.
- (e) T/F: LOOCV is very expensive for smoothing splines since we have such flexible models.
- (f) T/F: Local regression requires half of the training data each time we wish to make a prediction.
- (g) T/F: The span in local regression plays the same role as λ in smoothing splines.

- 4. State whether each of the below is true or false, and explain your answer.
- (a) T/F: We can determine the variance (and hence quantify uncertainty) at a given point when using an additive model
- (b) T/F: The tail behavior of splines makes them a good alternative to polynomial models
- (c) T/F: Step functions do not allow for interaction effects
- (d) T/F: Splines have large amounts of continuity compared to other methods we know
- (e) T/F: A cubic spline has more constraints than a natural cubic spline
- (f) T/F: There is no close form expression to determine where to place knots
- (g) T/F: The limiting behavior as lambda approaches infinity, for a smoother spline, is a linear function
- (h) T/F: The coefficients of GAMs are highly interpretable

- 5. Consider the nonlinear models we have discussed in class:
- a. Why do we say that we are not really interested in the coefficients of our nonlinear terms, and instead focus on the fitted function values?
- b. How might we determine the degree of a polynomial variable or the number of knots in a spline?
- c. What benefit do splines provide over piecewise polynomials? Why might we choose splines based on this?
- d. What benefit do splines provide over high degree piecewise functions? Why might we choose splines based on this?
- e. Since splines can have high variance at the outer range of the predictors, how can we account for this?

6. Consider GAMs:

- a. What is the benefit of the additivity of GAMs?
- b. When might a GAM outperform a more complex model?
- c. When might a GAM outperform a less complex model?
- d. What is the benefit of using a GAM versus using a single spline? What is the cost?
- e. What are the strengths and limitations of GAMs?

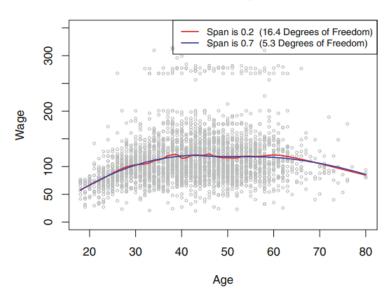
7. Smoothing Splines: answer the following questions and explain your answers. g is a smoothing spline if it minimizes

$$\sum_{i=1}^{n} (y_i - g(x_i))^2 + \lambda \int g''(t)^2 dt.$$

- (a) What does the first derivative of the function g measure?
- (b) What does the second derivative of the function g measure?
- (c) Conceptually, what is the $\int g''(t)^2 dt$ term doing?
- (d) As λ increases, what happens to the smoothness of q?
- (e) What happens when $\lambda = 0$? When $\lambda = \infty$?
- (f) Conceptually, what is λ controlling and how?

8. Explain how the span s controls the local linear regression fit in this plot. How would you choose the span s in practice?

Local Linear Regression



9. ISL Chapter 7, Exercise 1

10. ISL Chapter 7, Exercise 5