Python Companion to ISLR

Naresh Gurbuxani

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1 Introduction

Figure 1 shows graphs of Wage versus three variables.

Figure 2 shows boxplots of previous days' percentage changes in S&P 500 grouped according to today's change ${\tt Up}$ or ${\tt Down}.$

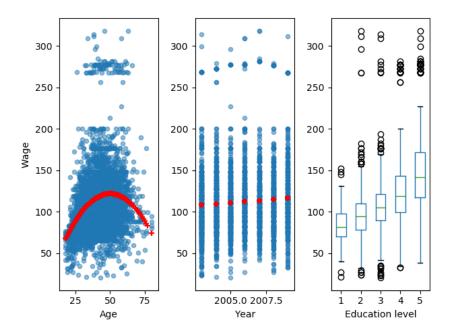


Figure 1: Wage data, which contains income survey information for males from the central Atlantic region of the United States. Left: wage as a function of age. On average, wage increases with age until about 60 years of age, at which point it begins to decline. Center: wage as a function of year. There is a slow but steady increase of approximately \$10,000 in the average wage between 2003 and 2009. Right: Boxplots displaying wage as a function of education, with 1 indicating the lowest level (no highschool diploma) and 5 the highest level (an advanced graduate degree). On average, wage increases with the level of education.

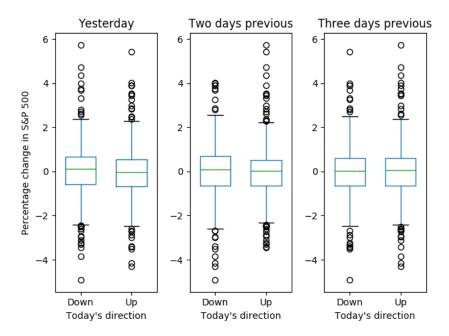


Figure 2: Left: Boxplots of the previous day's percentage change in the S&P 500 index for the days for which the market increased or decreased, obtained from the Smarket data. Center and Right: Same as left panel, but the percentage changes for two and three days previous are shown.

2 Statistical Learning

Figure 3 shows scatter plots of sales versus TV, radio, and newspaper advertising. In each panel, the figure also includes an OLS regression line.

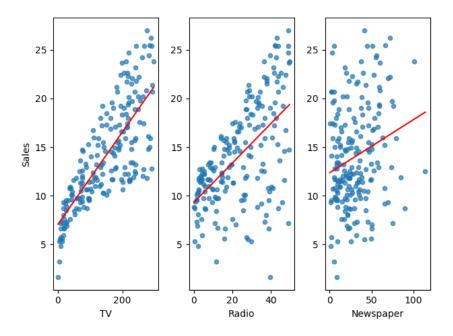


Figure 3: The Advertising data set. The plot displays sales, in thousands of units, as a function of TV, radio, and newspaper budgets, in thousands of dollars, for 200 different markets. In each plot we show the simple least squares fit of sales to that variable. In other words, each red line represents a simple model that can be used to predict sales using TV, radio, and newspaper, respectively.

Figure 4 is a plot of Income versus Years of Education from the Income data set. In the left panel, the "true" function (given by blue line) is actually my guess.

Figure 5 is a plot of Income versus Years of Education and Seniority from the Income data set. Since the book does not provide the true values of Income, "true" values shown in the plot are actually third order polynomial fit.

Figure 6 shows an example of the parametric approach applied to the Income data from previous figure.

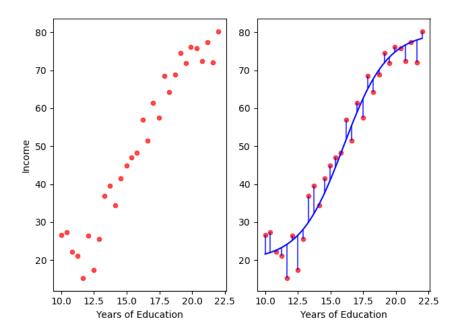


Figure 4: The Income data set. Left: The red dots are the observed values of income (in tens of thousands of dollars) and years of education for 30 individuals. Right: The blue curve represents the true underlying relationship between income and years of education, which is generally unknown (but is known in this case because the data are simulated). The vertical lines represent the error associated with each observation. Note that some of the errors are positive (when an observation lies above the blue curve) and some are negative (when an observation lies below the curve). Overall, these errors have approximately mean zero.

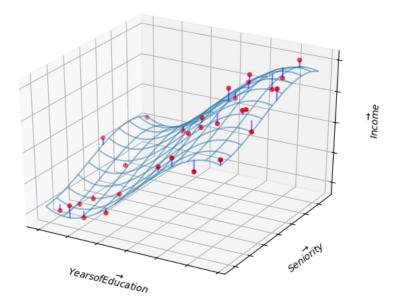


Figure 5: The plot displays income as a function of years of education and seniority in the Income data set. The blue surface represents the true underlying relationship between income and years of education and seniority, which is known since the data are simulated. The red dots indicate the observed values of these quantities for 30 individuals.

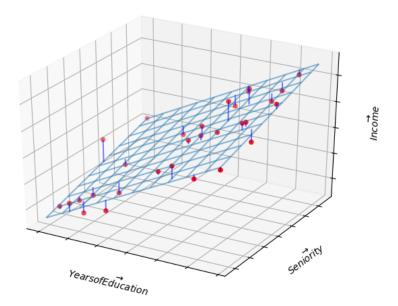


Figure 6: A linear model fit by least squares to the Income data from figure 5. The observations are shown in red, and the blue plane indicates the least squares fit to the data.

Figure 7 provides an illustration of the trade-off between flexibility and interpretability for some of the methods covered in this book.

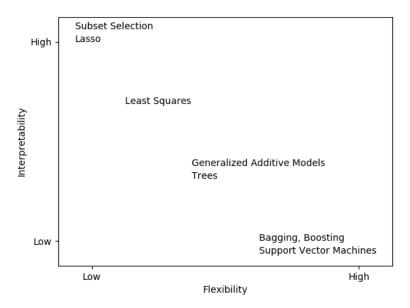


Figure 7: A representation of the tradeoff between flexibility and interpretability, using different statistical learning methods. In general, as the flexibility of a method increases, its interpretability decreases.