

Consumer Price Index Forecast with Demographic and Economic Leading Indicators

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1 Consumer Price Index Forecast Variables and Methods

Following is a brief analysis, to forecast the Consumer Price Index, as a measure of inflation. It is not meant to be a comprehensive, broad treatment of the many variables that might affect inflation. Rather, the analysis focuses on some economic trends that have been in the news lately, and that are sometimes said to portend a possible increase in inflation. For example, the U.S. national debt to GDP has rarely been higher, and has caused great concern for many. Also, the rate of inflation has just begun to increase, after years of being at record lows. Lastly among economic variables, the velocity of money is sometimes said to be related to inflation, as well. This analysis examines these variables, and their combined relationship to inflation, as measured by CPI. The data in this study goes back to 1961, and extends through 2017.

In addition to the economic variables, demographic variables were also considered. As a proxy for age distributions, the number of births by year, and also the birth rate by year, were included along with the previously mentioned economic variables. It is well known that the population of the U.S. has been aging. As older, retired individuals tend to purchase fewer consumer goods, this has helped inflation to remain very low.

The difference between the number of people entering age 65, the standard retirement age, compared to those entering age 45, the age many reach their peak earnings growth potential, can provide a useful indicator as to the net impact of age demographics on inflation. This variable was also included in this analysis, to determine how demographic forces interact with the other key economic variables in this forecast.

2 Arima Univariate Forecast of the Consumer Price Index

Before examining the combined, multivariate impact of all the variables on changes in the Consumer Price Index, this section will perform a univariate analysis to create a forecast on nothing more than the CPI data alone. The method utilized is an auto-ARIMA, or

automatic Auto Regressive Integrated Moving Average. This brief document will not go into the details as to how the ARIMA method works. However, perhaps it will suffice to say that it examines the following.

- Auto Regressive trends - the AR in ARIMA: measures a trends correlation to itself, such as with seasonality, or in the case of annual data, possible cyclicalilty
- Integration - the I in ARIMA: measures the degree to which subtracting each preceding time period's value removes the overall trend, also known as detrending.
- Moving Average - the MA in ARIMA: measures the extent to which a trend can be explained by a moving average pattern

The `auto.arima` function in R estimates each of these ARIMA parameters, and uses them to calculate a forecast. The plot in Figure 1 shows that this method estimates that the year to year trend in the CPI will be declining for the next five years, beginning in 2018.

As an ARIMA forecast is univariate, this method does not take into account any of the previously mentioned economic and demographic variables. It only assumes that past patterns in the trend of the CPI will continue. In the next section, we will evaluate the potential impact of the previously mentioned economic and demographic variables.

3 Multivariate Forecast with BVAR

In this section, we produce a multivariate forecast of the CPI, utilizing the previously mentioned economic and demographic variables. Older methods of performing multivariate time series analysis involved searching for top correlations between the predictor variables, and the target variable we are trying to predict. It largely ignored the impact of predictor variables on each other. A newer method for accounting for these cross variable impacts, is called Bayesian Vector Auto Regression, or BVAR.

Figure 2 shows a forecast of the CPI using the BVAR method. The black trend line to the right of the blue, dashed, vertical line, indicates the forecasted trend in the CPI for five years, after 2017. The forecast suggests that annual CPI will largely increase on the whole.

4 Influencers

Figure 3 shows the top variables in the BVAR forecast model, as to their influence on each other. Recall that the BVAR method takes into account not only each variable's impact on the target variable, the CPI in this case, but also on each other. The BVAR also examines the relationships between the variables, adjusting for many possible lags of the data over time. This enables finding the variables that tend to be leading indicators. This is a form of causality analysis, which investigates whether the strongest relationship, akin to the concept of correlation, occurs when the movement of one variable, precedes a corresponding movement on the part of another.

We can see in Figure 3 that all the predictor variables tend to be relatively strong influencers of the velocity of M2 money supply, as shown in the variable, `m2_velocity`. (M2

Forecasts from ARIMA(1,2,1)

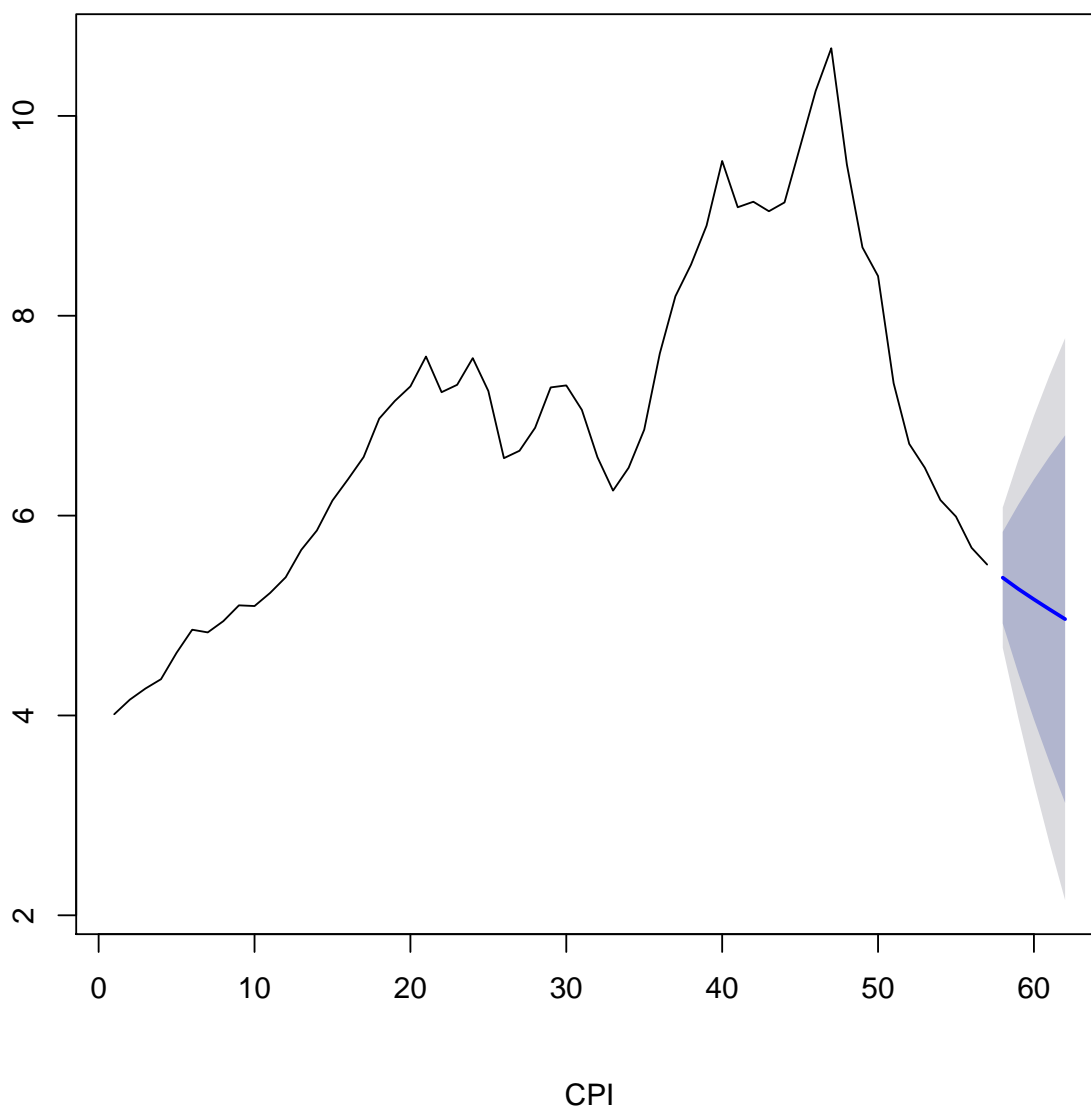


Figure 1: Univariate ARIMA forecast of the Consumer Price Index

money supply includes the cash and deposits represented by M1 money supply, plus what is known as near money, which includes such things as savings and money market accounts, and other cash deposits.) However, the predictor variables are not as strong as influencers of the Consumer Price Index. There may be other variables that are more predictive, and could increase forecast confidence.

It might be surprising to some, that the debt to GDP ratio does not appear as a more prominent influencer, given all the media attention the national debt has received. However, there have been other years in history that saw worse debt to GDP ratios, and it did not

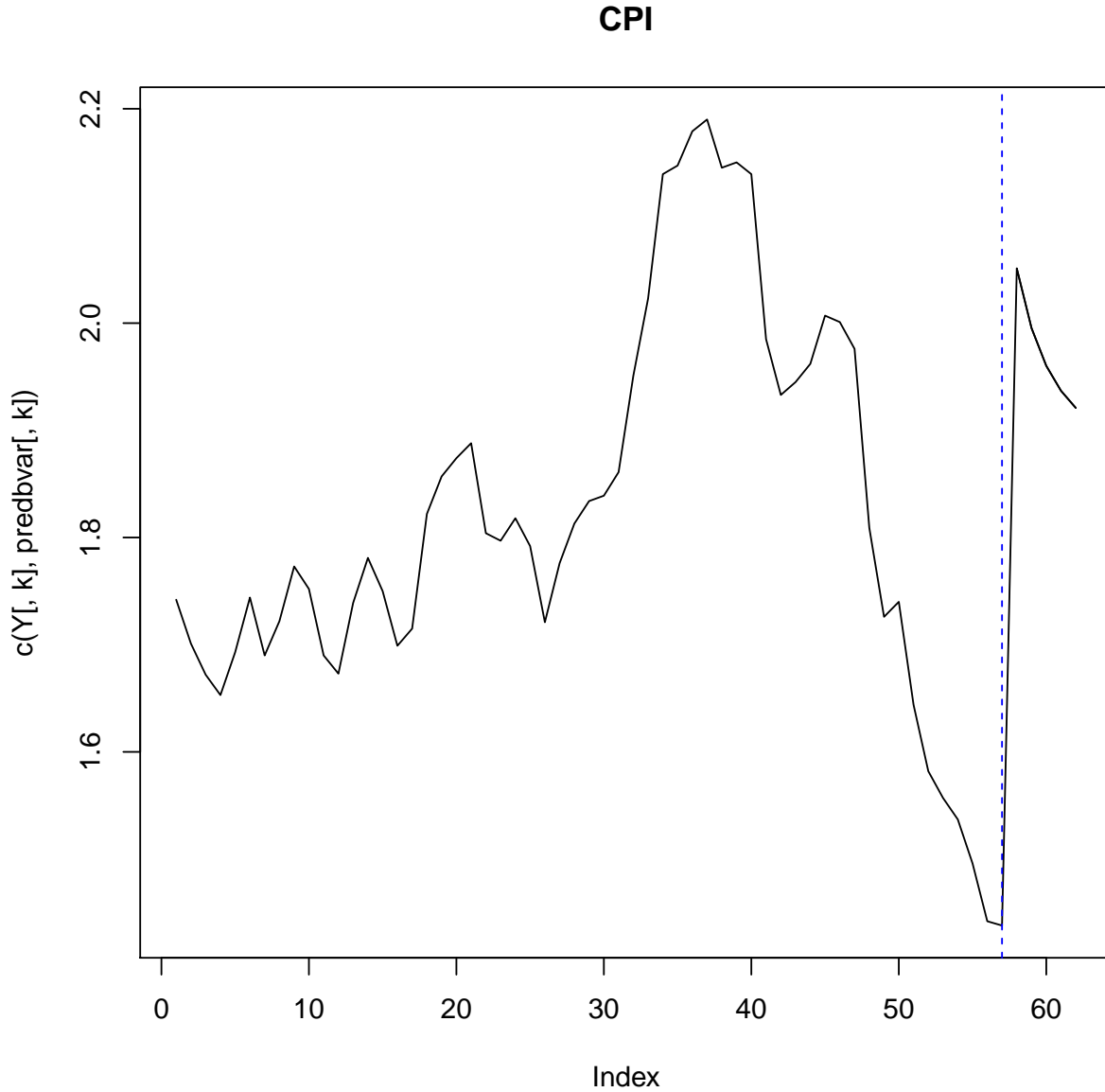


Figure 2: Forecast of the Consumer Price Index, utilizing predictor variables that are scaled, and pre-whitened - Bayesian Vector Auto Regression (BVAR)

necessarily drive high inflation. Also, while some have suggested that the millennial generation may influence inflation upwards as they begin to set up households, and increase demand for all the goods and services that go with that endeavor, it does not appear that this is a primary influencer of inflation this time around. It may be that the Baby Boomer generation is still a primary counter-acting factor.

Since the velocity of M2 money supply is so affected by the other predictor variables, it might be interesting to see how a forecast of M2 might be affected. It is easy to imagine how

inflation, demographics, and debt factors would affect the velocity of money moving in and out of the highly liquid accounts represented by M2 money supply. It is not certain whether the direction of velocity would move upward or downward in such a forecast, however.

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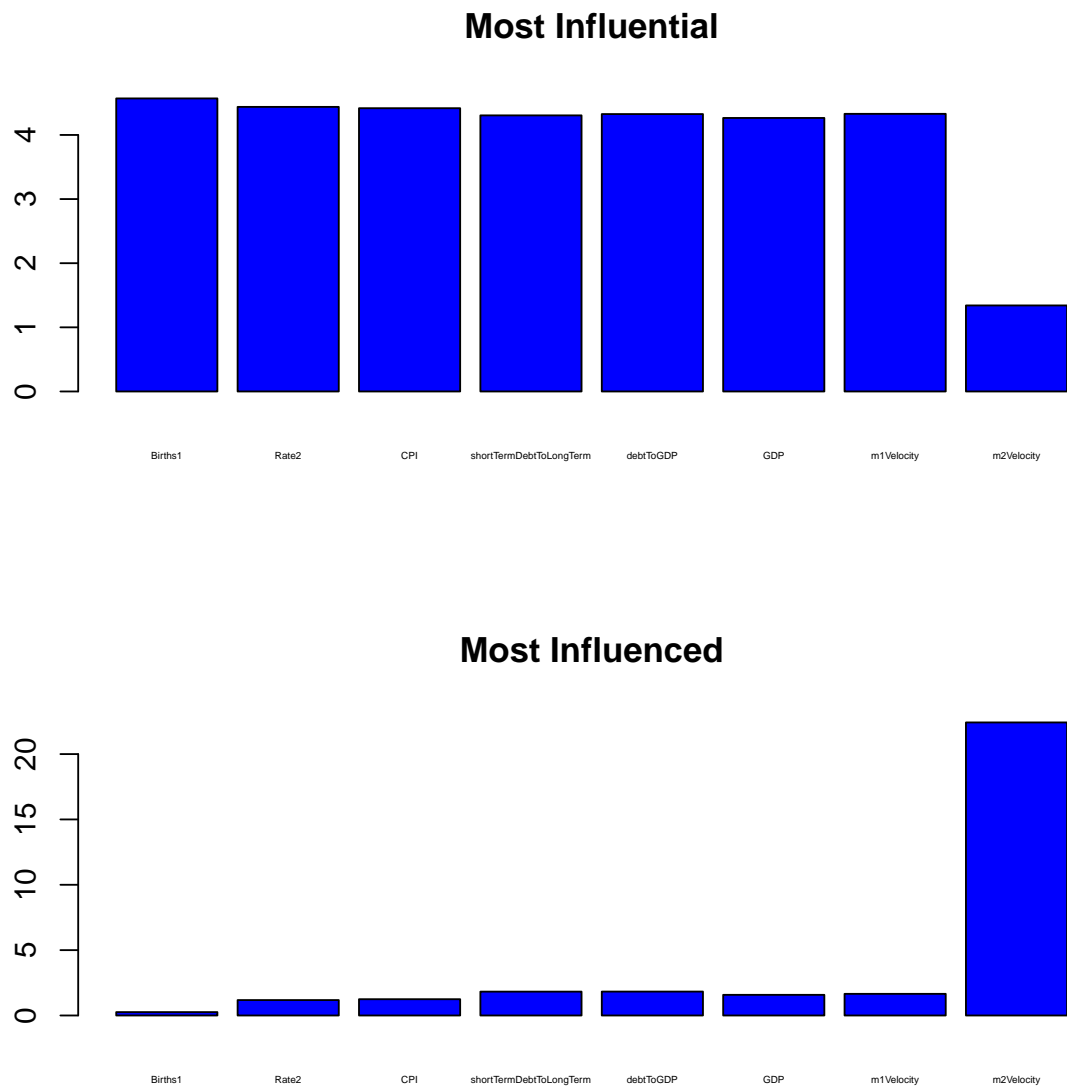


Figure 3: Main influencers of the Consumer Price Index - Bayesian Vector Auto Regression (BVAR)

5 Summary

This analysis examined the impact of some economic and demographic trends that have gotten attention in the media, as to their influence on the inflation. Many have predicted an increase in inflation, as a result of some of these trends. While no analysis can guarantee against such a collapse, and there are many other variables involved with market performance than these few, this analysis suggests that recent trends in these variables, when taken together, do indeed point to the possibility of increasing inflation.