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# FORECASTING

## PRINCIPLES AND PRACTICE

A comprehensive introduction to the latest forecasting methods using R. Learn to improve your forecast accuracy using dozens of real data examples.



3RD EDITION

 **OTexts**  
Open Texts Publishing

## 7. Time series regression models

### 7.1 The linear model

[OTexts.org/fpp3/](https://OTexts.org/fpp3/)

# Multiple regression and forecasting

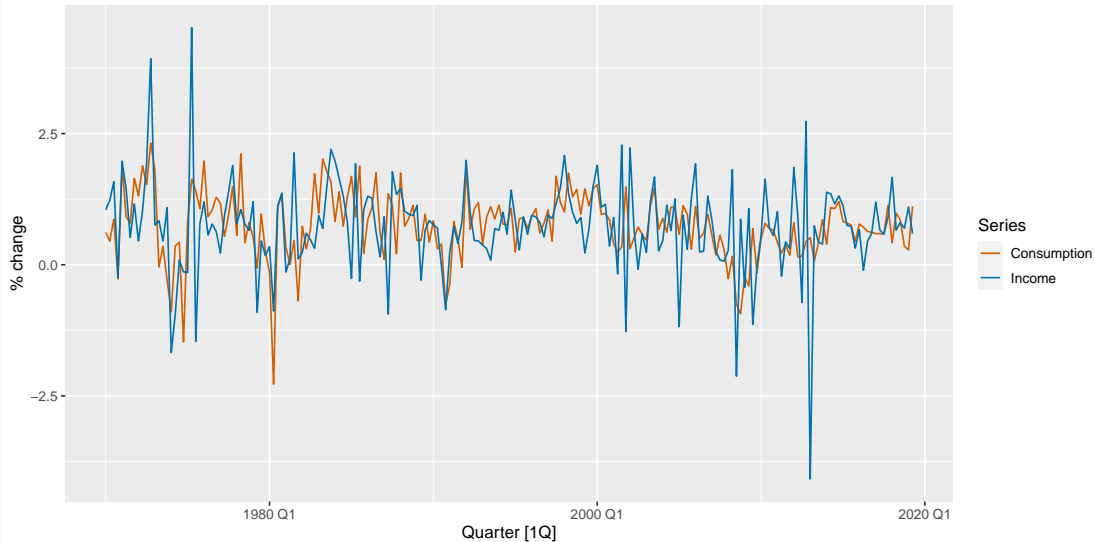
$$y_t = \beta_0 + \beta_1 x_{1,t} + \beta_2 x_{2,t} + \cdots + \beta_k x_{k,t} + \varepsilon_t.$$

- $y_t$  is the variable we want to predict: the “response” variable
- Each  $x_{j,t}$  is numerical and is called a “predictor”. They are usually assumed to be known for all past and future times.
- The coefficients  $\beta_1, \dots, \beta_k$  measure the effect of each predictor after taking account of the effect of all other predictors in the model.

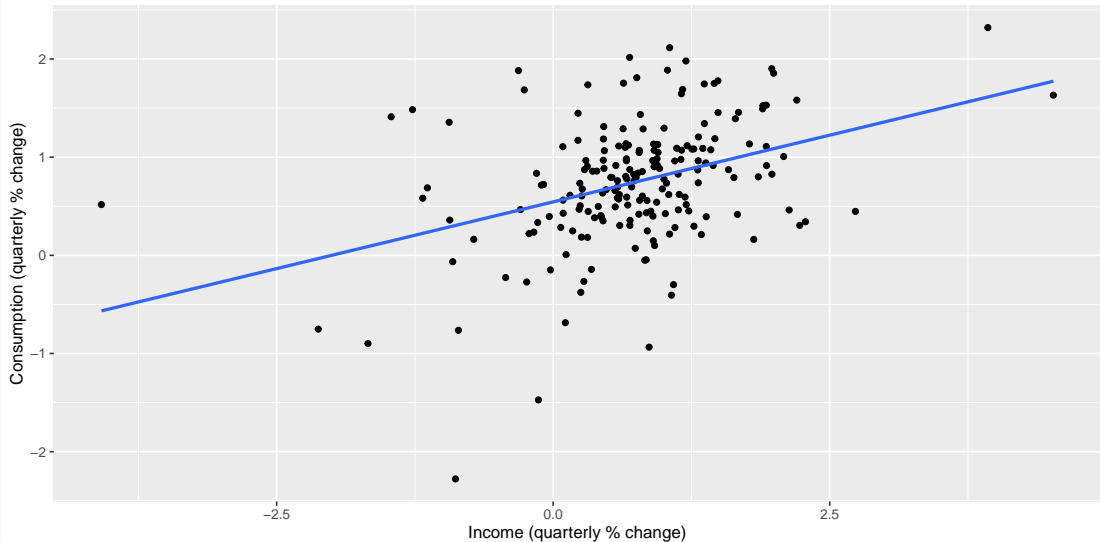
That is, the coefficients measure the **marginal effects**.

- $\varepsilon_t$  is a white noise error term

# Example: US consumption expenditure



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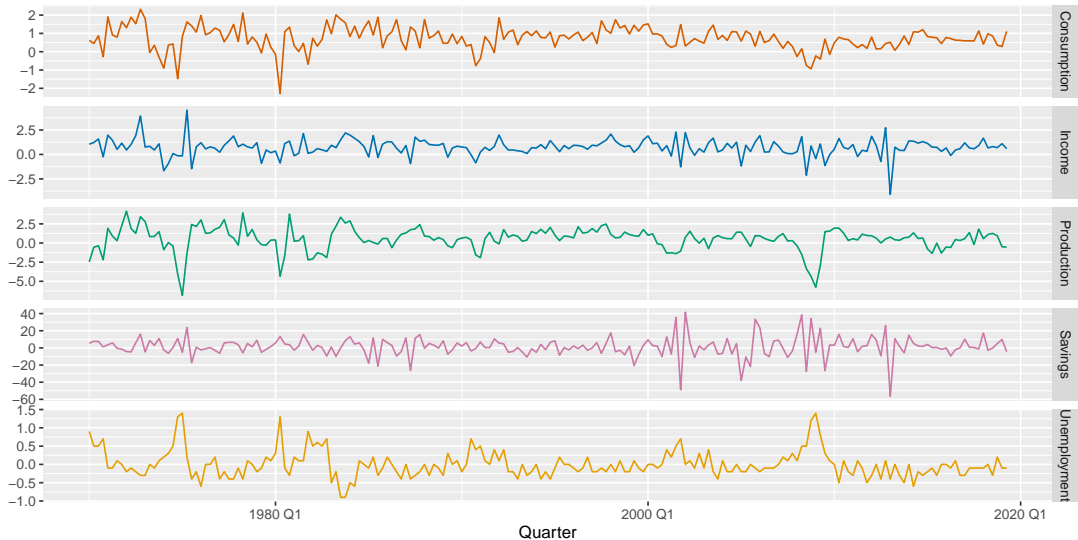


# Example: US consumption expenditure

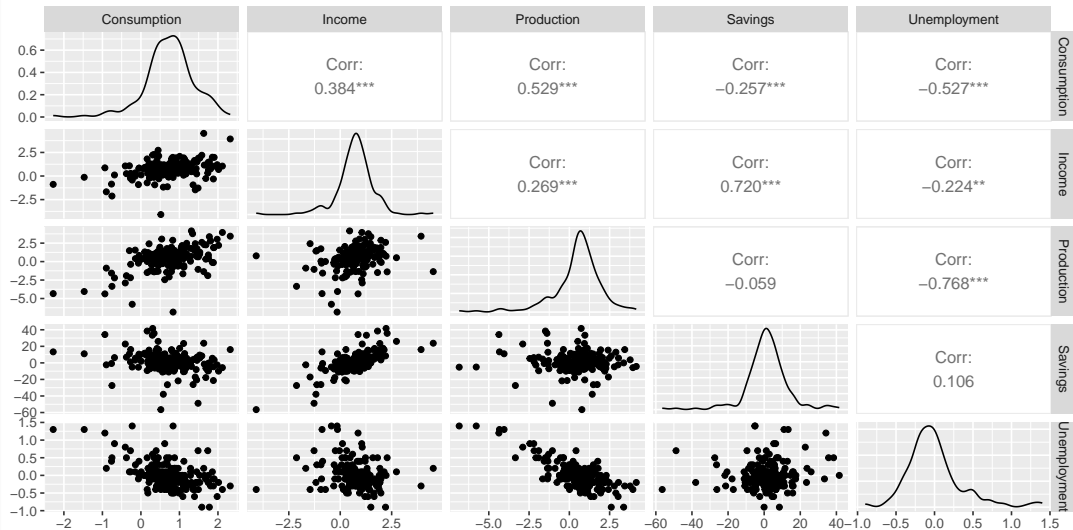
```
fit_cons <- us_change %>%  
  model(lm = TSLM(Consumption ~ Income))  
report(fit_cons)
```

```
## Series: Consumption  
## Model: TSLM  
##  
## Residuals:  
##      Min      1Q  Median      3Q      Max  
## -2.582 -0.278  0.019  0.323  1.422  
##  
## Coefficients:  
##              Estimate Std. Error t value Pr(>|t|)  
## (Intercept)   0.5445     0.0540   10.08 < 2e-16 ***  
## Income        0.2718     0.0467    5.82  2.4e-08 ***  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 0.591 on 196 degrees of freedom  
## Multiple R-squared:  0.147,    Adjusted R-squared:  0.143  
## F-statistic: 33.8 on 1 and 196 DF, p-value: 2e-08
```

# Example: US consumption expenditure



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# Assumptions for the linear model

For forecasting purposes, we require the following assumptions:

- $\varepsilon_t$  have mean zero and are uncorrelated.
- $\varepsilon_t$  are uncorrelated with each  $x_{j,t}$ .



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It is **useful** to also have  $\varepsilon_t \sim N(0, \sigma^2)$  when producing prediction intervals or doing statistical tests.