Rob J Hyndman George Athanasopoulos

# FORECASTING PRINCIPLES AND PRACTICE

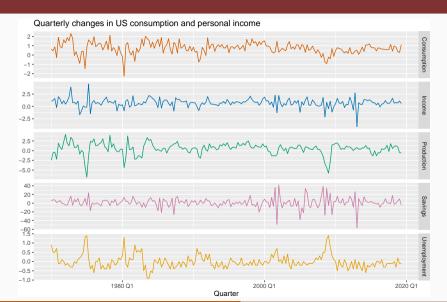


# 10. Dynamic regression models

10.2 Dynamic regression using fable OTexts.org/fpp3/

#### Regression with ARIMA errors

- In fable, we can specify an ARIMA(p, d, q) for the errors, and d levels of differencing will be applied to all variables  $(y, x_{1,t}, \ldots, x_{k,t})$  during estimation.
- Check that  $\varepsilon_t$  series looks like white noise.
- AICc can be calculated for final model.
- Repeat procedure for all subsets of predictors to be considered, and select model with lowest AICc value.

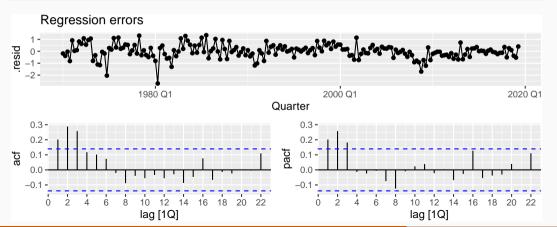


```
fit <- us_change |> model(ARIMA(Consumption ~ Income))
report(fit)
## Series: Consumption
## Model: LM w/ ARIMA(1,0,2) errors
##
## Coefficients:
##
          ar1
                 mal ma2 Income intercept
  0.707 -0.617 0.2066 0.1976
##
                                        0.595
## s.e. 0.107 0.122 0.0741 0.0462
                                        0.085
##
## sigma^2 estimated as 0.3113: log likelihood=-163
## AIC=338 AICc=339 BIC=358
```

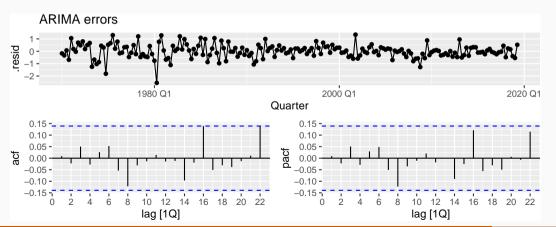
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```

Write down the equations for the fitted model.

```
residuals(fit, type = "regression") |>
    gg_tsdisplay(.resid, plot_type = "partial") +
    labs(title = "Regression errors")
```



```
residuals(fit, type = "innovation") |>
    gg_tsdisplay(.resid, plot_type = "partial") +
    labs(title = "ARIMA errors")
```



## 1 ARIMA(Consumption ~ Income) 5.54

0.785