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FORECASTING PRINCIPLES AND PRACTICE

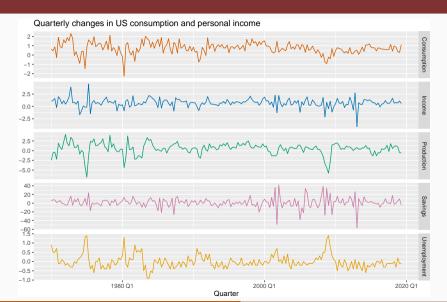


10. Dynamic regression models

10.2 Dynamic regression using fable OTexts.org/fpp3/

Regression with ARIMA errors

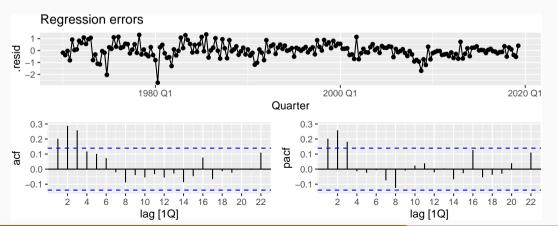
- In fable, we can specify an ARIMA(p, d, q) for the errors, and d levels of differencing will be applied to all variables $(y, x_{1,t}, \ldots, x_{k,t})$ during estimation.
- Check that ε_t series looks like white noise.
- AICc can be calculated for final model.
- Repeat procedure for all subsets of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



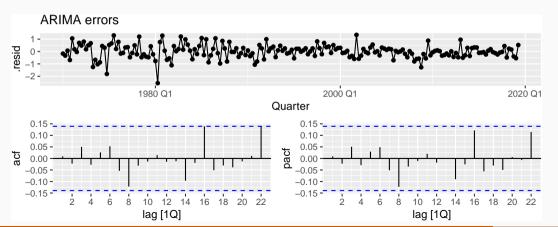
```
fit <- us change |> model(ARIMA(Consumption ~ Income))
report(fit)
## Series: Consumption
## Model: LM w/ ARIMA(1,0,2) errors
##
## Coefficients:
##
          ar1
                 mal ma2 Income intercept
##
  0.707 -0.617 0.2066 0.1976
                                        0.595
## s.e. 0.107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

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```

```
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