

# **Introdução à Inferência Bayesiana**

Démerson André Polli

ENAP - 16/01/2020

# Petrobras (PETR4)

```
petr4 = c(24.16, 24.49, 24.72, 24.85, 24.85, 24.85, 24.46, 24.75, 24.86, 24.11, 24.11, 24.11,  
24.11, 24.69, 24.93, 24.90, 25.11, 25.33, 25.33, 25.33, 25.35, 24.81, 24.42, 24.46, 24.18,  
24.18, 24.18, 25.03, 25.35, 26.23, 26.12, 26.04, 26.04, 26.04, 26.65, 26.32, 26.67, 26.40,  
25.99, 25.99, 25.99, 25.87, 26.36, 26.34, 25.99, 26.04, 26.04, 26.04, 26.04, 26.04, 26.05,  
25.96, 26.97, 26.97, 26.97, 26.75, 27.25, 27.37, 27.46, 28.02, 28.02, 28.02, 28.42, 28.19,  
27.80, 26.28, 26.61, 26.61, 26.61, 27.86, 26.61, 27.31, 27.31, 27.25, 27.25, 27.25, 27.53,  
26.80, 27.71, 28.01, 28.47, 28.47, 28.47, 28.38, 28.01, 27.25, 25.14, 25.24, 25.24, 25.24,  
26.00, 26.03, 26.86, 26.71, 26.71, 26.71, 26.71, 26.94, 26.86, 27.08, 26.52, 26.64, 26.64,  
26.64, 26.38, 26.01, 26.01, 26.13, 26.05, 26.05, 26.05, 25.64, 26.64, 26.11, 25.97, 25.21,  
25.21, 25.21, 25.30, 25.19, 24.59, 24.02, 24.84, 24.84, 24.84, 25.78, 25.66, 25.22, 25.47,  
25.62, 25.62, 25.62, 26.16, 25.87, 25.53, 24.94, 25.37, 25.37, 25.37, 25.58, 25.24, 25.65,  
26.12, 26.02, 26.02, 26.02, 26.51, 26.21, 26.53, 26.42, 26.46, 26.46, 26.46, 26.80, 26.86,  
27.61, 27.61, 27.58, 27.58, 27.58, 26.85, 27.01, 26.58, 26.76, 26.61, 26.61, 26.61, 26.18,  
26.48, 26.74, 26.75, 26.99, 26.99, 26.99, 27.40, 27.40, 27.72, 27.85, 27.51, 27.51, 27.51,  
27.17, 27.05, 26.85, 26.79, 26.84, 26.84, 26.84, 26.86, 26.70, 26.25, 25.52, 25.75, 25.75,  
25.75, 25.62, 25.46, 24.99, 25.89, 24.94, 24.94, 24.94, 25.26, 24.99, 25.70, 25.65, 25.04,  
25.04, 25.04, 25.35, 24.49, 23.81, 23.50, 23.62, 23.62, 23.62, 23.61, 25.01, 24.79, 23.86,  
23.55, 23.55, 23.55, 23.92, 24.17, 25.06, 25.06, 24.86, 24.86, 24.86, 25.16, 25.81, 25.94,  
26.06, 26.47, 26.47, 26.47, 26.63, 26.41, 26.59, 26.42, 27.58, 27.58, 27.58, 27.21, 26.75,  
26.82, 26.53, 27.01, 27.01, 27.01, 26.80, 26.87, 27.22, 27.18, 27.08, 27.08, 27.08, 27.04,  
26.26, 26.28, 26.05, 25.72, 25.72, 25.72, 25.57, 26.06, 26.28, 26.79, 26.84, 26.84, 26.84,  
27.12, 27.45, 27.18, 27.12, 27.29, 27.29, 27.29, 28.08, 28.45, 27.83, 28.75, 29.09, 29.09,  
29.09, 29.31, 29.56, 29.87, 29.91, 29.84, 29.84, 29.84, 29.14, 29.20, 30.37, 29.50, 29.93,  
29.93, 29.93, 29.67, 29.55, 28.96, 28.74, 28.74, 28.74, 28.74, 28.44, 29.50, 29.50, 29.63,  
29.38, 29.38, 29.38, 28.85, 28.99, 29.18, 28.81, 28.73, 28.73, 28.73, 28.64, 29.31, 29.70,  
29.99, 29.86, 29.86, 29.86, 30.08, 30.04, 30.61, 29.63, 29.07, 29.07, 29.07, 29.50, 30.18,  
30.25, 29.91, 30.14, 30.14, 30.14, 30.55, 30.55, 30.55, 30.52, 30.18, 30.18, 30.18, 30.70,  
30.70, 30.70, 30.45, 30.81, 30.81, 30.81, 30.69, 30.50, 30.40, 30.27, 30.33, 30.33, 30.33,  
30.00, 29.55)
```

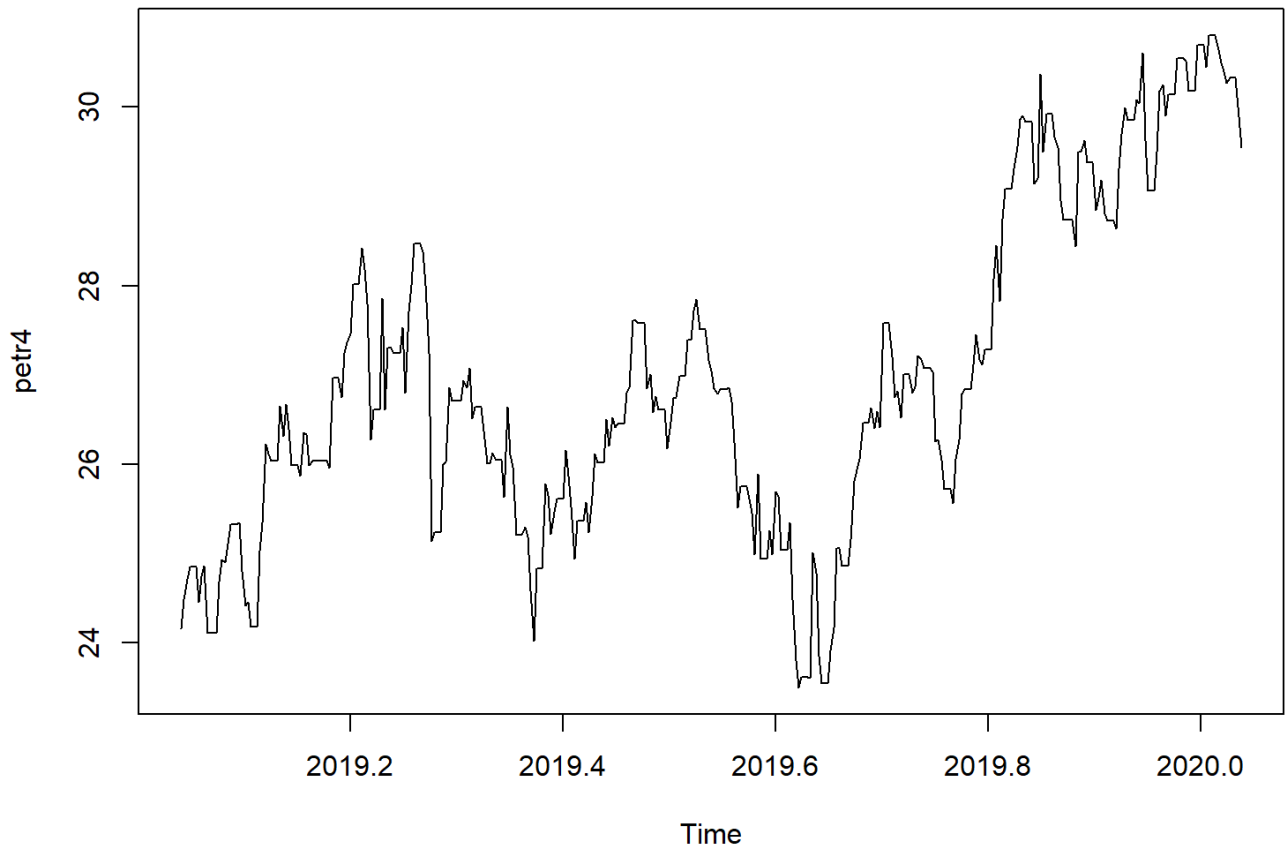
# Petrobras (PETR4) - configurando a série

- Série de cotações da PETR4 (fechamento) entre 16 de janeiro de 2019 e 15 de janeiro de 2020.

```
inds <- seq(as.Date("2019-01-16"), as.Date("2020-01-15"), by = "day")

## Create a time series object
petr4 <- ts(petr4,
            start = c(2019, as.numeric(format(inds[1], "%j"))),
            frequency = 365)
```

# Petrobras (PETR4) - série



# Braskem (BRKM3)

```
brkm3 = c(47.02,45.46,44.01,44.01,44.01,44.40,46.35,47.36,47.51,47.51,47.51,47.51,46.87,  
47.17,50.49,50.33,50.33,50.33,50.33,51.11,52.09,52.36,50.14,50.14,50.14,50.14,  
49.87,51.21,51.62,53.47,52.59,52.59,52.59,53.53,53.54,54.02,53.14,53.65,53.65,  
53.65,53.74,54.99,53.08,52.48,53.54,53.54,53.54,53.54,53.54,53.50,52.09,52.08,  
52.08,52.08,52.69,52.62,52.62,53.10,53.10,53.10,53.10,52.96,51.62,51.62,49.95,  
49.95,49.95,50.43,51.12,48.68,51.16,51.63,51.63,51.63,50.14,48.67,48.67,  
48.09,46.83,46.83,46.83,47.22,47.69,47.43,47.70,49.04,49.04,49.04,46.97,46.25,  
46.19,46.34,46.34,46.34,46.34,45.79,45.27,44.79,44.93,44.64,44.64,44.64,45.74,  
46.61,46.61,44.72,44.67,44.67,44.67,43.44,42.32,40.90,39.07,38.75,38.75,38.75,  
36.51,37.00,37.67,36.02,35.44,35.44,35.44,38.65,40.69,39.91,40.11,39.62,39.62,  
39.62,38.56,38.94,38.58,38.06,38.94,38.94,38.94,37.90,33.56,33.76,33.34,32.83,  
32.83,32.83,32.67,33.10,33.10,33.94,34.13,34.13,34.13,33.78,34.08,34.08,34.08,  
34.75,34.75,34.75,34.87,34.26,34.08,33.59,34.07,34.07,34.07,34.40,34.18,34.40,  
35.16,36.46,36.46,36.46,36.54,36.54,36.02,36.07,36.41,36.41,36.41,36.13,35.29,  
35.24,34.95,34.90,34.90,34.90,34.75,35.34,35.53,34.57,34.17,34.17,34.17,33.68,  
34.06,33.92,33.15,32.34,32.34,32.34,32.03,32.51,32.51,31.97,31.79,31.79,31.79,  
31.47,31.41,30.76,29.55,29.40,29.40,29.40,29.69,28.62,27.84,27.60,27.62,27.62,  
27.62,27.06,27.16,27.90,28.81,28.82,28.82,28.82,29.08,28.77,29.95,30.32,30.57,  
30.57,30.57,31.25,30.72,30.81,30.22,30.07,30.07,30.07,29.69,30.37,30.71,31.10,  
31.37,31.37,31.37,32.43,32.09,31.54,32.13,32.04,32.04,32.04,32.76,32.03,30.84,  
30.91,30.98,30.98,30.98,29.75,29.00,28.75,29.50,29.78,29.78,29.78,30.10,30.44,  
29.81,29.81,29.50,29.50,29.50,29.15,29.30,29.11,29.15,28.69,28.69,28.69,29.00,  
28.94,29.19,28.88,28.94,28.94,28.94,29.50,29.36,30.16,30.10,30.51,30.51,30.51,  
31.81,31.02,30.50,30.49,30.49,30.49,30.49,29.50,28.90,28.90,29.00,29.35,29.35,  
29.35,29.10,29.00,29.05,29.00,28.61,28.61,28.61,28.70,28.99,29.20,28.86,29.50,  
29.50,29.50,29.93,29.60,29.62,29.72,29.20,29.20,29.20,31.49,30.15,30.17,30.00,  
30.00,30.00,30.00,30.10,30.10,30.10,30.60,31.38,31.38,31.38,31.54,31.54,31.54,  
32.75,33.46,33.46,33.46,33.87,34.09,34.99,36.50,37.85,37.85,37.85,36.10,35.82,  
35.00)
```

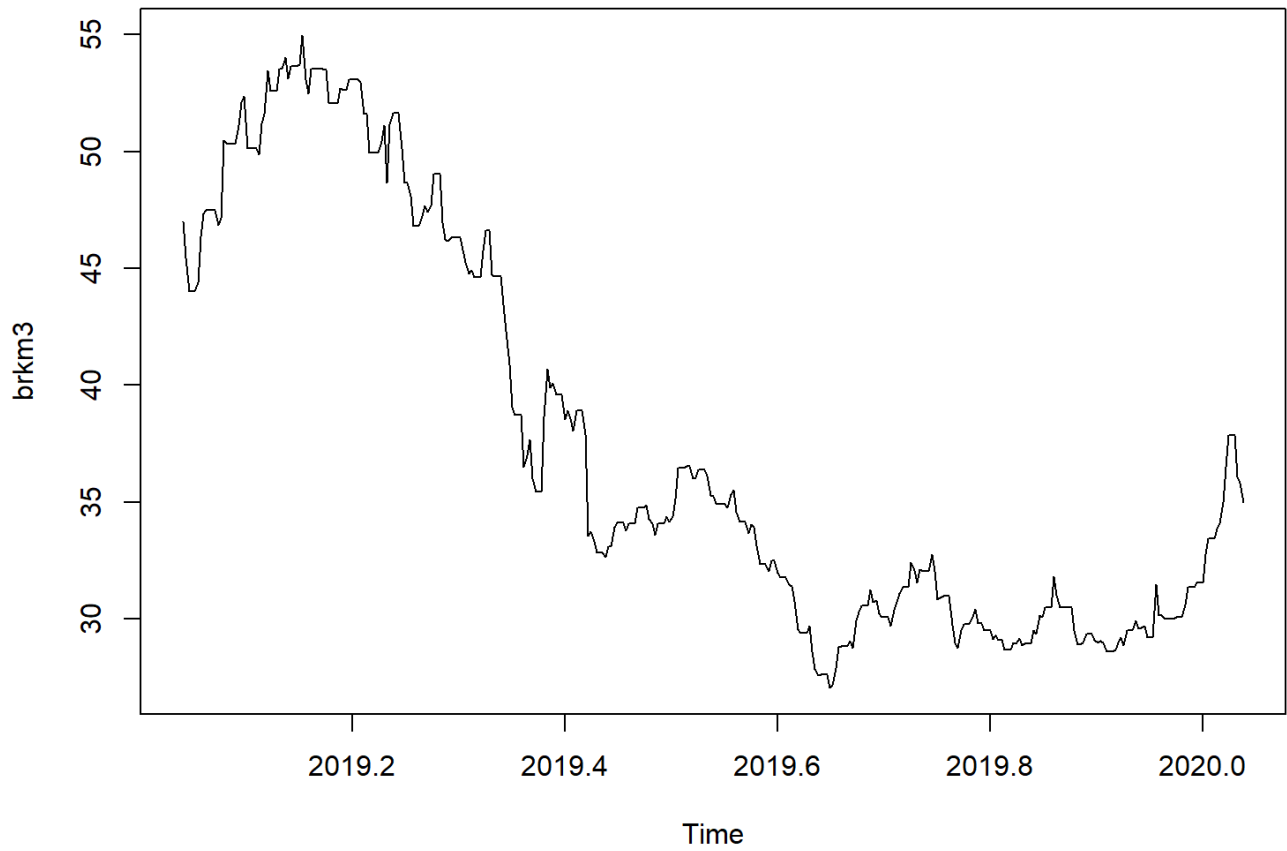
# Petrobras (BRKM3) - configurando a série

- Série de cotações da BRKM3 (fechamento) entre 16 de janeiro de 2019 e 15 de janeiro de 2020.

```
inds <- seq(as.Date("2019-01-16"), as.Date("2020-01-15"), by = "day")

## Create a time series object
brkm3 <- ts(brkm3,
            start = c(2019, as.numeric(format(inds[1], "%j"))),
            frequency = 365)
```

# Braskem (BRKM3) - série



# IBOV

```

ibov = c( 94.39, 95.35, 96.10, 96.10, 96.10, 96.01, 95.10, 96.56, 97.68, 97.68, 97.68,
          97.68, 95.44, 95.64, 97.00, 97.39, 97.86, 97.86, 97.86, 98.59, 98.31, 94.64,
          94.11, 95.34, 95.34, 95.34, 94.41, 96.17, 96.12, 98.02, 97.53, 97.53, 97.53,
          96.51, 97.66, 96.55, 96.93, 97.89, 97.89, 97.89, 97.24, 97.60, 97.31, 95.58,
          94.60, 94.60, 94.60, 94.60, 94.60, 94.22, 94.34, 95.37, 95.37, 95.37, 98.03,
          97.83, 98.90, 98.61, 99.14, 99.14, 99.14, 99.99, 99.59, 98.04, 96.73, 93.74,
          93.74, 93.74, 93.66, 95.31, 91.90, 94.39, 95.42, 95.42, 95.42, 96.05, 95.39,
          94.49, 96.31, 97.11, 97.11, 97.11, 97.37, 96.29, 95.95, 94.82, 92.88, 92.88,
          92.88, 93.08, 94.33, 93.29, 94.58, 94.58, 94.58, 94.58, 94.59, 95.92, 95.05,
          96.39, 96.24, 96.24, 96.24, 96.19, 96.35, 96.35, 95.53, 96.01, 96.01, 96.01,
          95.01, 94.02, 95.60, 94.81, 94.26, 94.26, 94.26, 91.73, 92.09, 91.62, 90.02,
          89.99, 89.99, 89.99, 91.95, 94.49, 94.36, 93.91, 93.63, 93.63, 93.63, 94.86,
          96.39, 96.57, 97.46, 97.03, 97.03, 97.03, 97.02, 97.38, 96.00, 97.21, 97.82,
          97.82, 97.82, 97.73, 98.96, 98.32, 98.77, 98.04, 98.04, 98.04, 97.62, 99.40,
          100.30,100.30,102.01,102.01,102.01,102.06,100.09,100.69,100.72,100.97,100.97,
          100.97,101.34,100.61,102.04,103.64,104.09,104.09,104.09,104.53,104.53,105.82,
          105.14,103.90,103.90,103.90,103.80,103.78,103.86,104.72,103.45,103.45,103.45,
          103.95,103.70,104.12,102.66,102.82,102.82,102.82,103.48,102.93,101.81,102.13,
          102.67,102.67,102.67,100.10,102.16,102.78,104.11,103.91,103.91,103.91,101.92,
          103.46,100.26, 99.06, 99.81, 99.81, 99.81, 99.47, 99.22,101.20,100.01, 97.67,
          97.67, 97.67, 96.43, 97.28, 98.19,100.52,101.14,101.14,101.14,100.63, 99.68,
          100.94,102.25,102.94,102.94,102.94,103.18,103.03,103.45,104.33,103.50,103.50,
          103.50,103.26,104.62,104.53,104.34,104.82,104.82,104.82,104.64,103.88,104.48,
          105.32,105.08,105.08,105.08,104.75,104.05,101.03,101.52,102.55,102.55,102.55,
          100.57, 99.98,101.25,101.82,103.83,103.83,103.83,104.30,104.49,105.42,105.02,
          104.73,104.73,104.73,106.02,107.38,107.54,106.99,107.36,107.36,107.36,108.19,
          107.56,108.41,107.22,108.20,108.20,108.20,108.65,108.72,108.35,109.58,107.63,
          107.63,107.63,108.37,106.75,106.06,106.56,106.56,106.56,106.56,106.27,105.86,
          105.86,107.50,108.69,108.69,108.69,108.42,107.06,107.71,108.29,108.23,108.23,
          108.23,108.93,108.96,110.30,110.62,111.13,111.13,111.13,110.98,110.67,110.96,
          112.14,112.57,112.57,112.57,111.90,112.62,114.32,115.13,115.12,115.12,115.12,
          115.07,115.07,115.07,117.20,116.53,116.53,116.53,115.65,115.65,115.65,118.57,
          117.78,117.78,117.78,116.59,116.66,116.17,115.95,115.50,115.50,115.50,117.33,
          117.63,116.41)

```



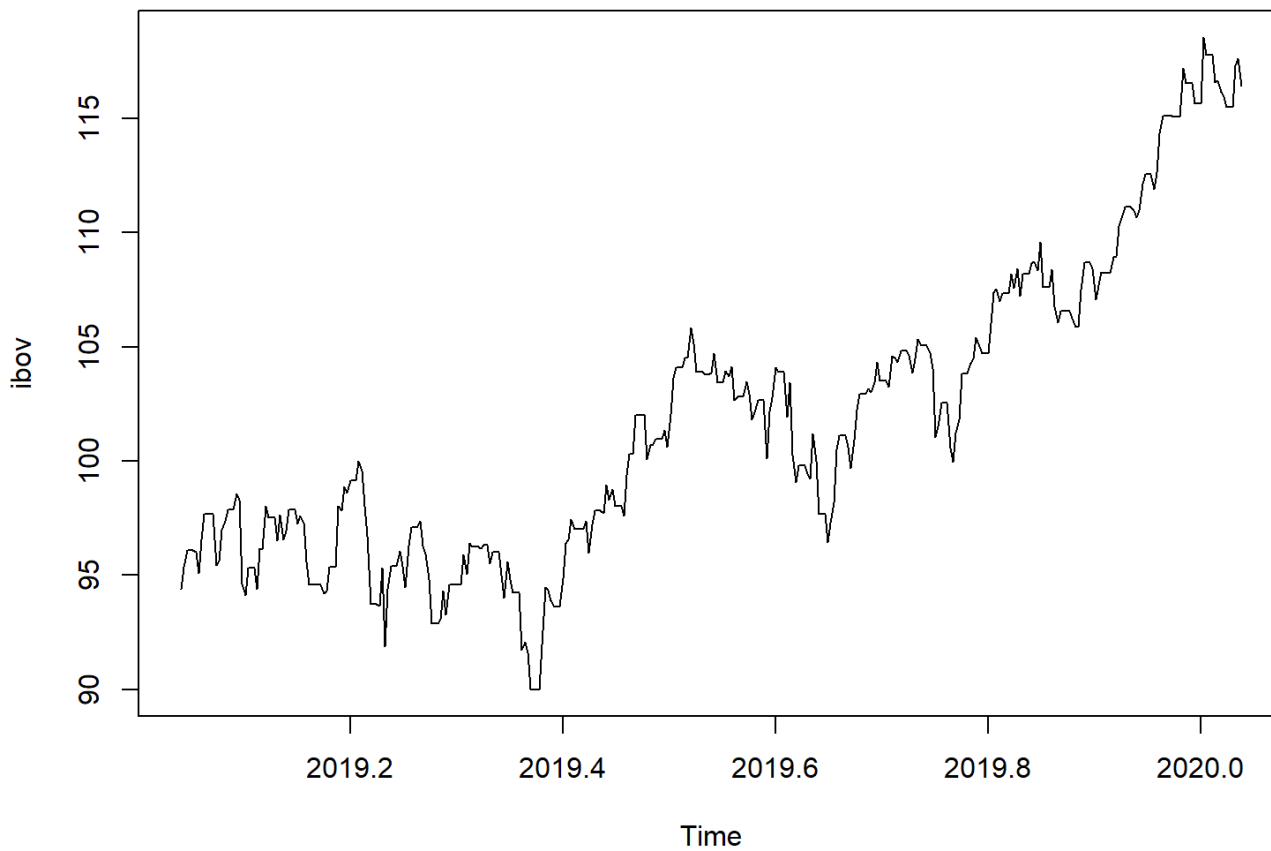
# IBOV - configurando a série

- Série de IBOV entre 16 de janeiro de 2019 e 15 de janeiro de 2020.

```
inds <- seq(as.Date("2019-01-16"), as.Date("2020-01-15"), by = "day")

## Create a time series object
ibov <- ts(ibov,
           start = c(2019, as.numeric(format(inds[1], "%j"))),
           frequency = 365)
```

# IBOVESPA - série



# Modelos de Espaço de Estados (I)

O modelo *Dynamic Linear Model* pode ser definido na forma

$$\mathbf{y}_t = \mathbf{F}_t \boldsymbol{\theta}_t + \boldsymbol{\nu}_t, \quad \boldsymbol{\nu}_t \sim N(\mathbf{0}, \mathbf{V}_t)$$

$$\boldsymbol{\theta}_t = \mathbf{G}_t \boldsymbol{\theta}_{t-1} + \boldsymbol{\omega}_t, \quad \boldsymbol{\omega}_t \sim N(\mathbf{0}, \mathbf{W}_t)$$

$$\boldsymbol{\theta}_0 \sim N(m_0, C_0)$$

- Estes modelos estão implementados no pacote `d1m` no R.
- O pacote `d1m` permite o ajuste por *máxima verossimilhança* ou *inferência bayesiana* de modelos de espaço de estados.
  - `dLmModARMA()`: processo ARMA
  - `dLmModPoLy()`: modelos polinomiais
  - `dLmModReg()`: regressão linear
  - `dLmModSeas()`: modelo sazonal (periódico)
  - `dLmModTrig()`: modelo sazonal (trigonométrico)

# Modelo de Espaço de estados (II)

```
if(!require("dlm")) install.packages("dlm")
if(!require("numDeriv")) install.packages("numDeriv")

library(dlm)
library(numDeriv)
```

# Modelo de Espaço de estados (III)

```
build = function(theta) {  
  dlmModPoly(order = 1, dV = theta[1], dW = theta[2])  
}  
  
(fit = dlmMLE(petr4, parm = c(100, 2), build, lower = c(rep(1e-4, 2))))
```

```
## $par  
## [1] 0.009248528 0.136705220  
##  
## $value  
## [1] -149.6691  
##  
## $counts  
## function gradient  
##      60      60  
##  
## $convergence  
## [1] 0  
##  
## $message  
## [1] "CONVERGENCE: REL_REDUCTION_OF_F <= FACTR*EPSMCH"
```

# Modelo de Espaço de estados (IV)

```
model <- build(fit$par)  
drop(V(model))
```

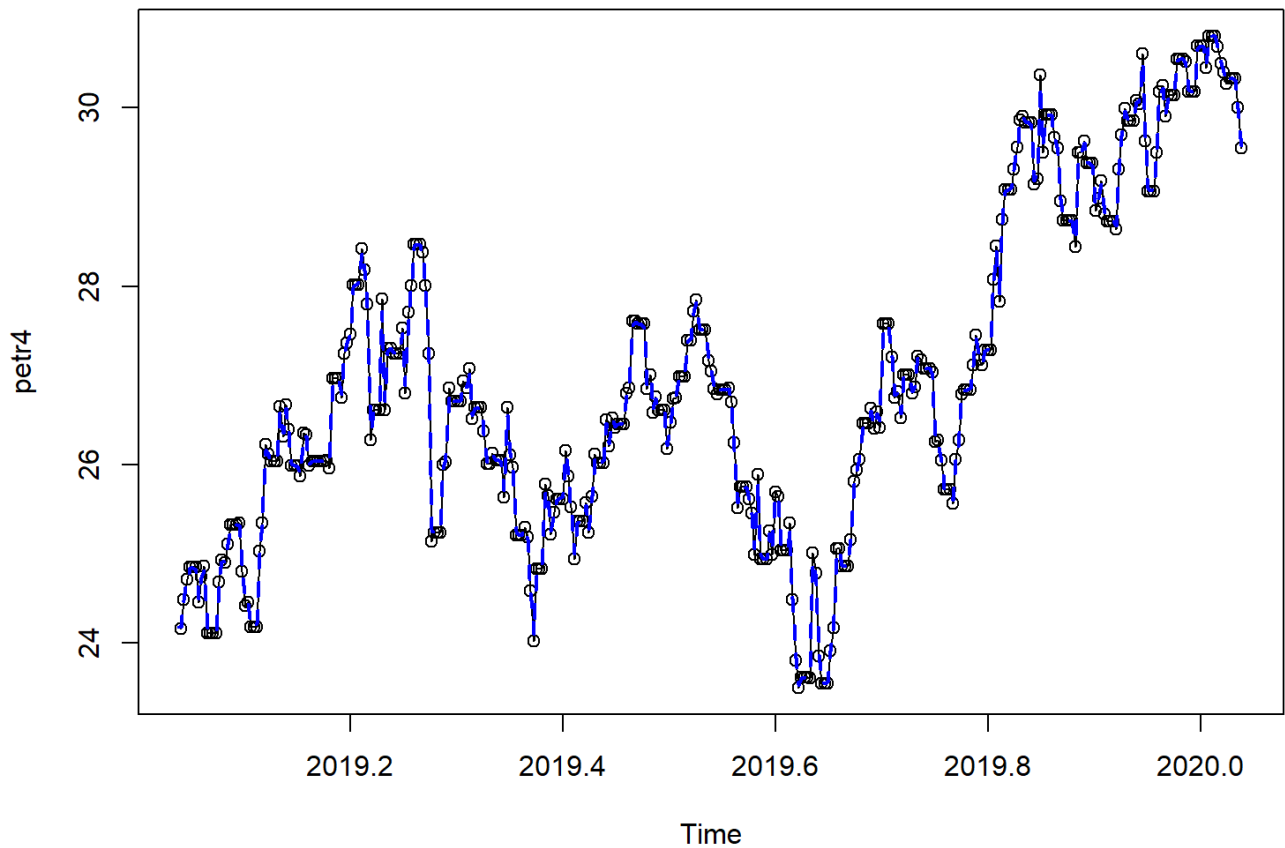
```
## [1] 0.009248528
```

```
drop(W(model))
```

```
## [1] 0.1367052
```

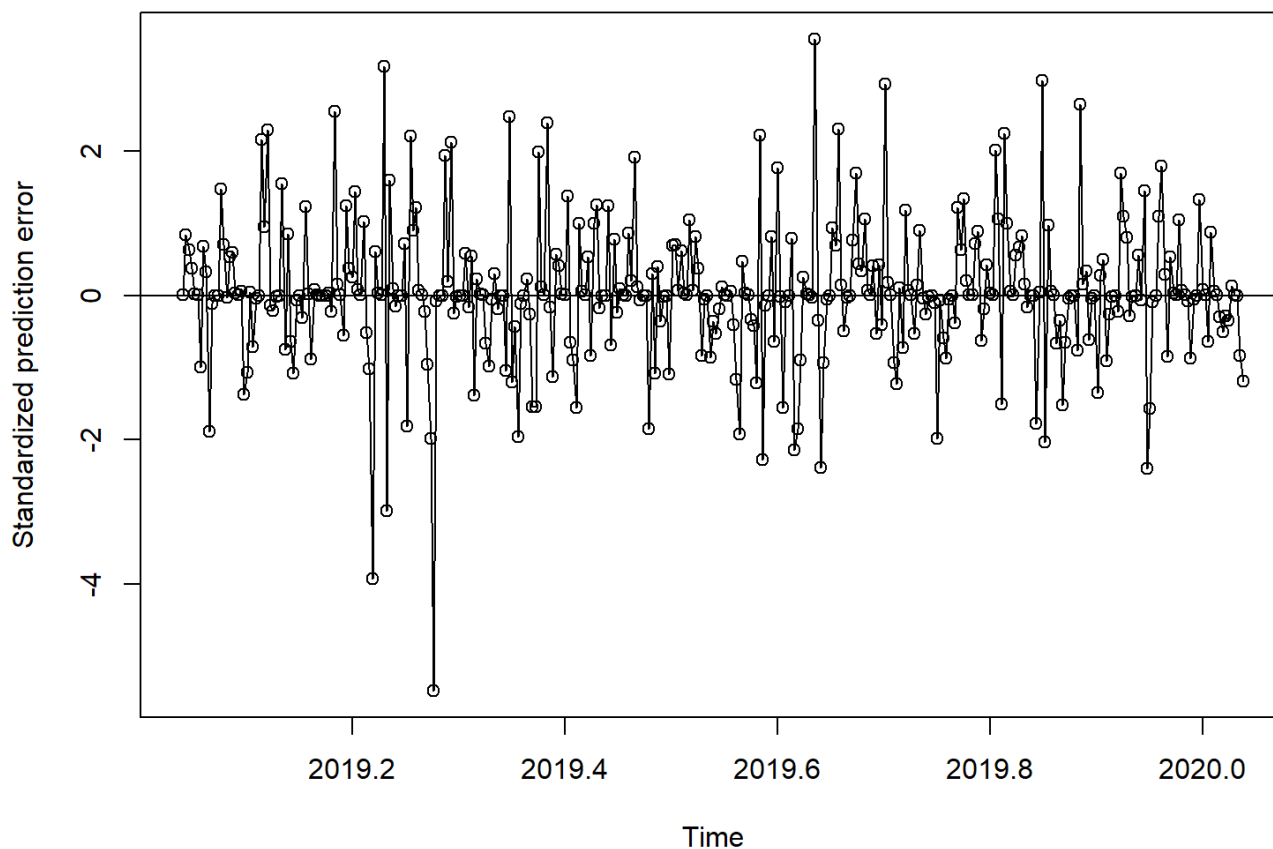
# Modelo de Espaço de estados (V)

```
smoothed <- dlmSmooth(petr4, model)
plot(petr4, type = "o")
lines(smoothed$s, lty = 2, col = "blue", lwd= 2)
```



# Modelo de Espaço de estados (V)

```
filtered <- dlmFilter(petr4, model)
plot(residuals(filtered, sd = FALSE), type = "o",
     ylab = "Standardized prediction error")
abline(h = 0)
```





# Modelo de Espaço de estados (VI)

```
build = function(theta) {  
  dlmModPoly(order = 2, dV = theta[1], dW = theta[2:3])  
}  
  
(fit = dlmMLE(petr4, parm = c(100, 2, 2), build, lower = c(rep(1e-4, 3))))
```

```
## $par  
## [1] 0.008886788 0.137671306 0.000100000  
##  
## $value  
## [1] -135.736  
##  
## $counts  
## function gradient  
##      88      88  
##  
## $convergence  
## [1] 52  
##  
## $message  
## [1] "ERROR: ABNORMAL_TERMINATION_IN_LNSRCH"
```

# Modelo de Espaço de estados (VII)

```
model <- build(fit$par)
drop(V(model))
```

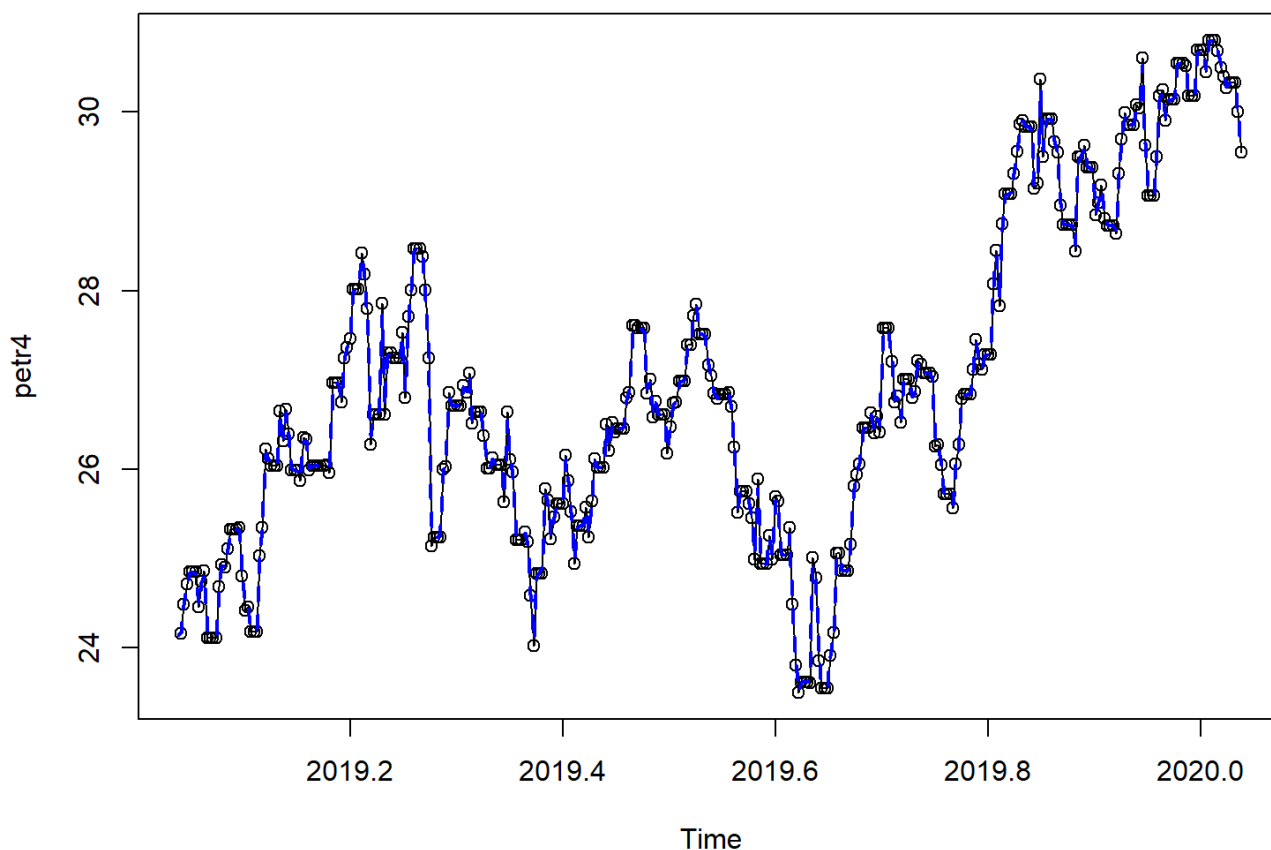
```
## [1] 0.008886788
```

```
drop(W(model))
```

```
##           [,1] [,2]
## [1,] 0.1376713 0e+00
## [2,] 0.0000000 1e-04
```

# Modelo de Espaço de estados (VIII)

```
smoothed <- dlmSmooth(petr4, model)
plot(petr4, type = "o")
lines(smoothed$s[,1], lty = 2, col = "blue", lwd = 2)
```

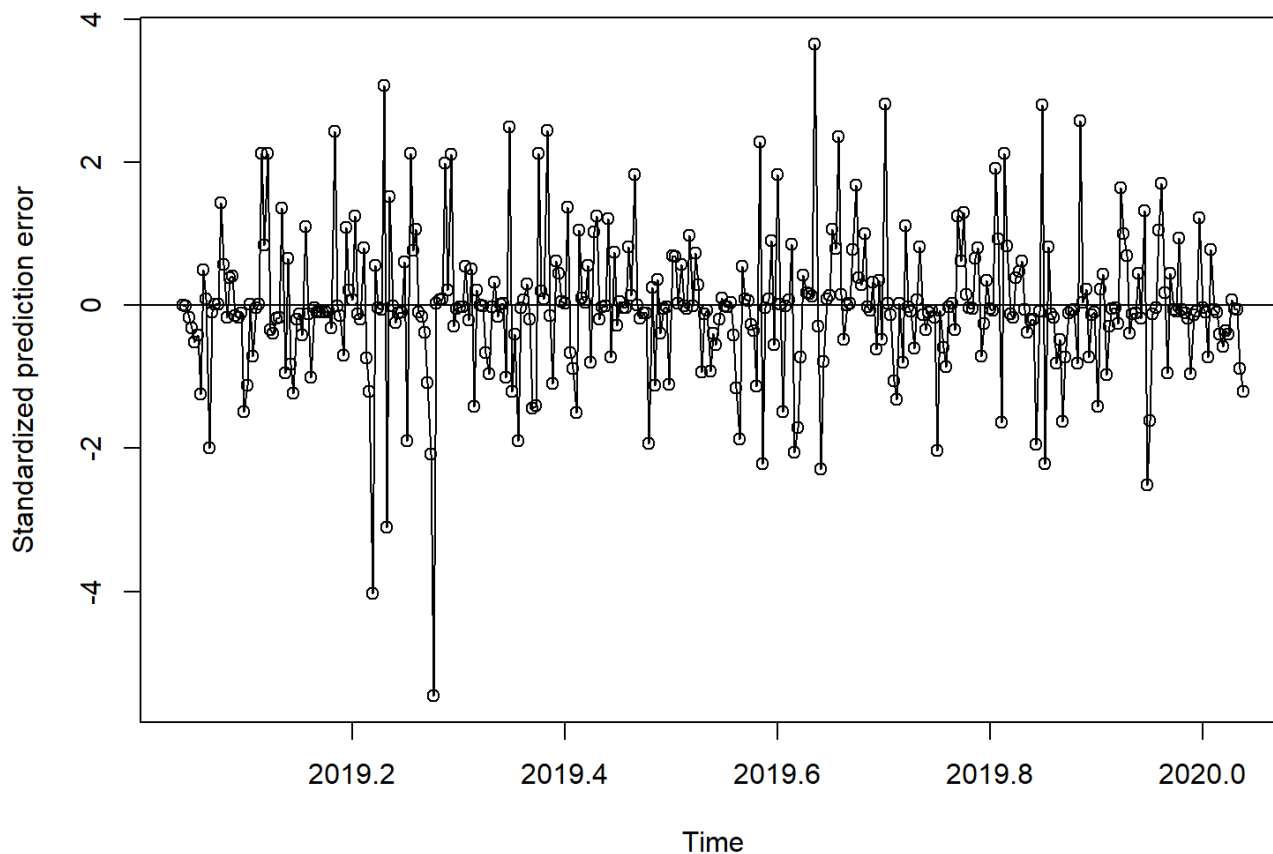


# Modelo de Espaço de estados (IX)

```
build = function(theta) {  
  dlmModPoly(order = 2, dV = theta[1], dW = theta[2:3]) +  
  dlmModARMA(ar = theta[4:5], ma = theta[6:7], sigma2 = theta[8])  
}  
  
fit = dlmMLE(petr4, parm = c(100, 2, 2, .1, .1, .1, .1, 1.5), build,  
  lower = c(rep(1e-4, 3), rep(-Inf, 4), 1e-4))
```

# Modelo de Espaço de estados (X)

```
filtered <- dlmFilter(petr4, model)
plot(residuals(filtered, sd = FALSE), type = "o",
     ylab = "Standardized prediction error")
abline(h = 0)
```

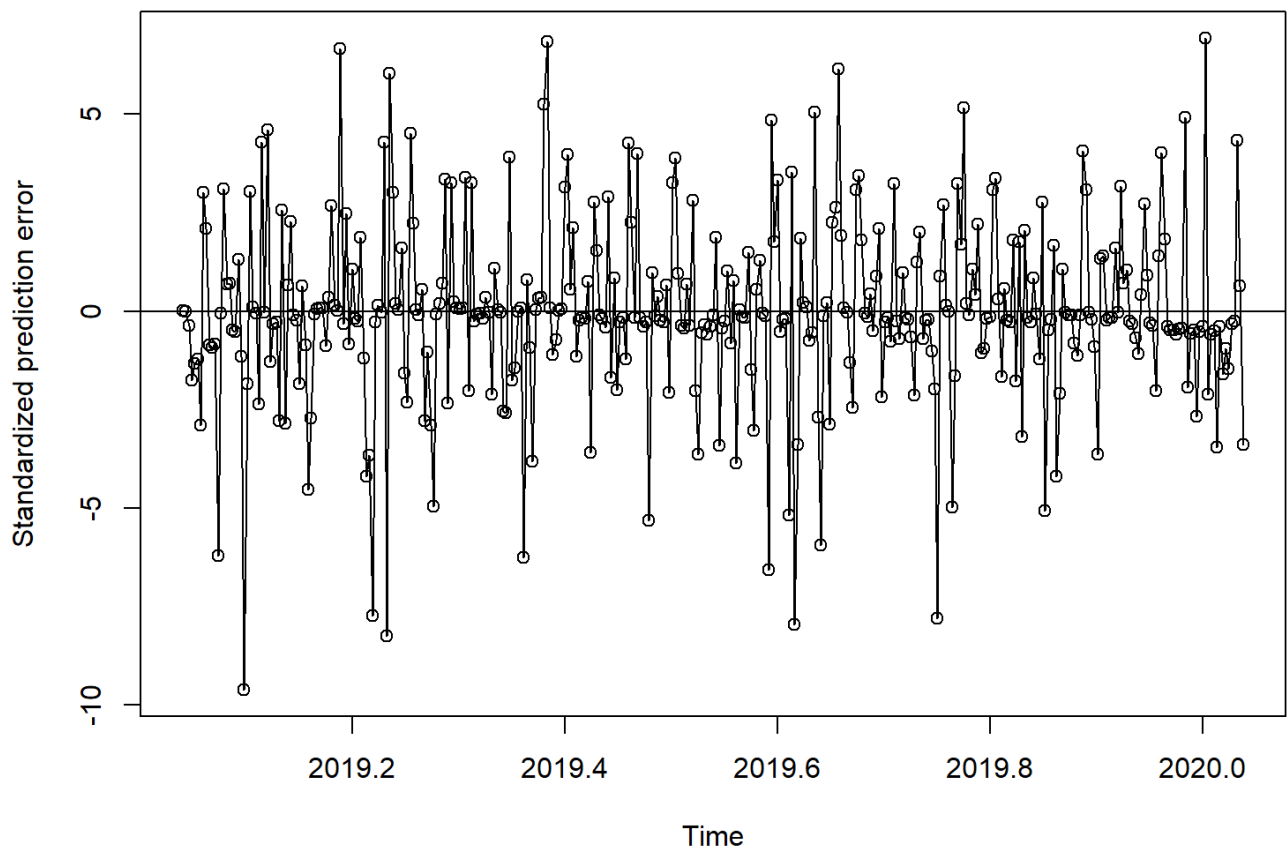


# Modelo de Espaço de estados (XI)

```
build = function(theta) {  
  dlmModPoly(order = 2, dV = theta[1], dW = theta[2:3]) +  
  dlmModARMA(ar = theta[4:5], ma = theta[6:7], sigma2 = theta[8]) +  
  dlmModReg(X = cbind(petr4, brkm3))  
}  
  
fit = dlmMLE(ibov, parm = c(100, 2, 2, .1, .1, .1, .1, 1.5), build,  
  lower = c(rep(1e-4, 3), rep(-Inf, 4), 1e-4))
```

# Modelo de Espaço de estados (XII)

```
filtered <- dlmFilter(ibov, model)
plot(residuals(filtered, sd = FALSE), type = "o",
     ylab = "Standardized prediction error")
abline(h = 0)
```



# Modelo de Espaço de estados (XIII)

```
build = function(theta) {  
  dlmModPoly(order = 1, dV = theta[1], dW = theta[2]) %+%  
  dlmModPoly(order = 1, dV = theta[1], dW = theta[2])  
}  
  
fit = dlmMLE(cbind(petr4, brkm3), parm = c(100, 2, 100, 2), build,  
            lower = rep(1e-4, 4))
```



# Modelo de Espaço de estados (X)

fit

```
## $par
## [1] 0.0001000 0.3406173 100.0000000 2.0000000
##
## $value
## [1] -11.71285
##
## $counts
## function gradient
##      28      28
##
## $convergence
## [1] 0
##
## $message
## [1] "CONVERGENCE: REL_REDUCTION_OF_F <= FACTR*EPSMCH"
```

# Próximos passos

- Como incluir séries de maior frequência como exógena?
- Como modelar séries multivariadas correlacionadas?
- Aplicação aos dados econométricos.