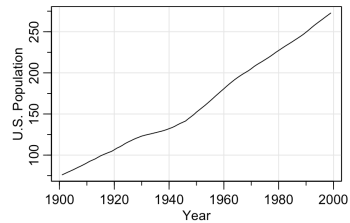


Time Series Cheat Sheet

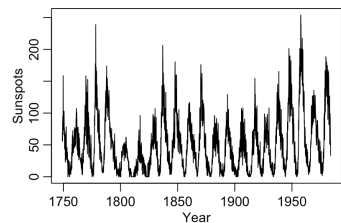


Plot Time Series

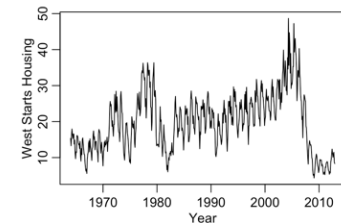
1. `tsplot(x=time, y=data)`



2. `plot(ts(data, start=start_time, frequency=gap))`



3. `ts.plot(ts(data, start=start_time, frequency=gap))`



Simulation

Autoregression of Order p

$$X_t = \phi_1 X_{t-1} + \phi_2 X_{t-2} + \dots + \phi_p X_{t-p} + W_t$$

Moving Average of Order q

$$X_t = Z_t + \theta_1 Z_{t-1} + \theta_2 Z_{t-2} + \dots + \theta_q Z_{t-p}$$

ARMA (p, q)

$$X_t = \phi_1 X_{t-1} + \phi_2 X_{t-2} + \dots + \phi_p X_{t-p} + Z_t + \theta_1 Z_{t-1} + \theta_2 Z_{t-2} + \dots + \theta_q Z_{t-p}$$

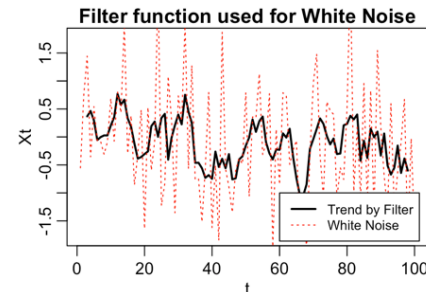
Simulation of ARMA (p, q)

`arima.sim(model=list(ar=c(ϕ_1, \dots, ϕ_p),
ma=c($\theta_1, \dots, \theta_q$)), n=n)`

Filters

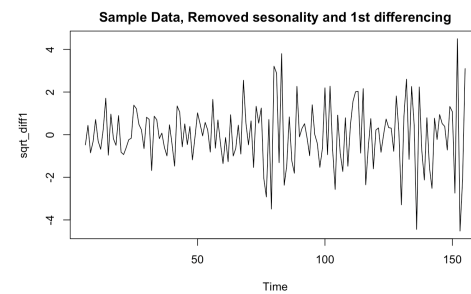
Linear Filter: filter()

`filter(data, filter=filter_coefficients, sides=2,
method="convolution", circular=F)`



Differencing Filter: diff()

`diff(data, lag=4, differences=1)`

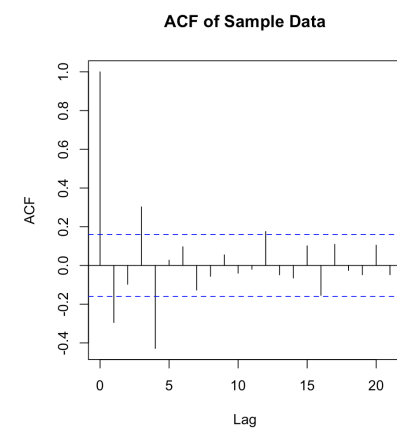


Auto-correlation

Use ACF and PACF to detect model

(Complete) Auto-correlation function: acf()

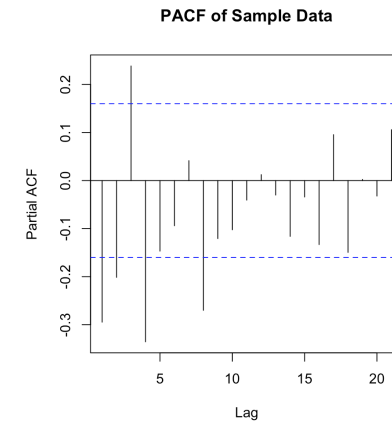
`acf(data, type='correlation', na.action=na.pass)`



Partial Auto-correlation function: pacf()

`pacf(data, na.action=na.pass)`

OR: `acf(data, type='partial', na.action=na.pass)`



Parameter Estimation

Fit an ARMA time series model to the data

ar(): To estimate parameters of an AR model

`ar(x=data, aic=T, order.max = NULL,
c("yule-walker", "burg", "ols", "mle", "yw"))`

Call:
`ar(x = sqrt1, aic = TRUE, order.max = NULL, method = c("yule-walker", "burg", "ols", "mle", "yw"))`
Coefficients:
1 2 3 4 5 6 7 8 9
-0.3066 -0.1903 0.0793 -0.5065 -0.1873 -0.1149 -0.0580 -0.3031 -0.1207
Order selected 9 sigma^2 estimated as 1.52

arima(): To estimate parameters of an AM or ARMA model, and build model

`arima(data, order=c(p, o, q),method=c('ML'))`

Call:
`arima(x = sqrt1, order = c(2, 0, 6), method = c("ML"))`
Coefficients:
ar1 ar2 ma1 ma2 ma3 ma4 ma5 ma6 intercept
-0.1658 -0.7951 -0.1536 0.7524 -0.2101 -0.7680 0.0605 -0.6812 -0.0048
s.e. 0.0918 0.0754 0.0916 0.1047 0.0794 0.0743 0.0812 0.0895 0.0062
sigma^2 estimated as 1.193: log likelihood = -230.78, aic = 481.55

AICc(): Compare models using AICC

`AICc(fittedModel)`

Forecasting

Forecasting future observations given a fitted ARMA model

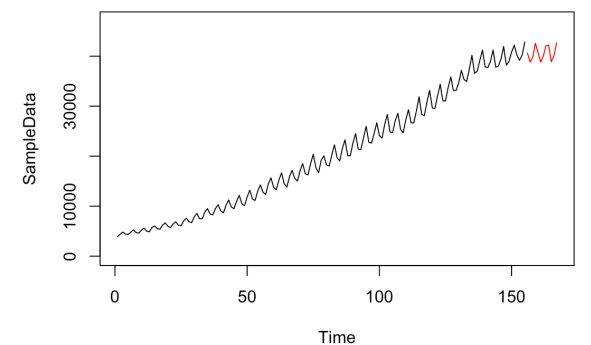
predict(): Predict future observations given a fitted ARMA model

`predict(arima_model, number_to_predict)`

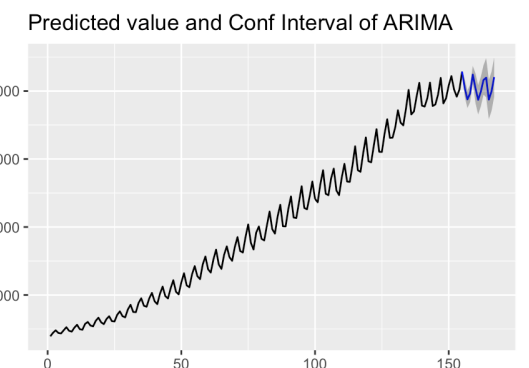
Plot Predicted values and Confidence Interval:

`fit<-predict(arima_model, number_to_predict)`

`ts.plot(data,
xlim=c(1, length(data)+number_to_predict),
ylim=c(0, max(fit$pred+1.96*fit$se)))
lines(length(data)+1:length(data)+
number_to_predict, fit$pred)`



OR: `autoplot(forecast(arima_model, level=c(95),
h=number_to_predict))`



Class Agnostic Time Series with tsbox :: CHEAT SHEET



Basics

IDEA

tsbox provides a time series toolkit which:

1. works identically with most time series **classes**
2. handles regular and irregular **frequencies**
3. **converts** between classes and frequencies

Most functions in tsbox have the same structure:

function starts with `ts_`

first argument is any `ts-boxable` object

```
a <- ts_pc(AirPassengers)
```

returns a `ts-boxable` object of the same class as input

COMBINE TIME SERIES

collect time series of **all classes** and **frequencies** as multiple time series

```
ts_c(mdeaths, austres)
```

combine time series to a new, single time series (first series wins if overlapping)

```
ts_bind(mdeaths, austres)
```

like `ts_bind`, but extra- and retropolate, using growth rates

```
ts_chain(mdeaths, austres)
```

PLOT AND SUMMARIZE

Plot time series of **all classes** and **frequencies**

```
ts_plot(mdeaths, austres)
ts_ggplot(mdeaths, austres)
```

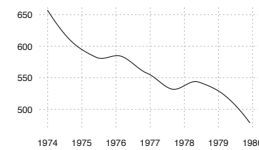
```
ts_summary(ts_c(mdeaths, austres))
```

	id	obs	diff	freq	start	end
1	mdeaths	72	1	month	1974-01-01	1979-12-01
2	austres	89	3	month	1971-04-01	1993-04-01

Helper Functions

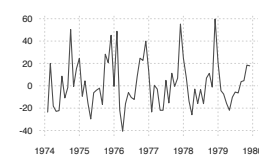
Transform time series of **all classes** and **frequencies**

TRANSFORM



ts_trend(): Trend estimation based on loess

```
ts_trend(fdeaths)
```



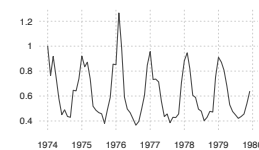
ts_pc(), ts_pcy(), ts_pca(), ts_diff(), ts_diffy(): (annualized) Percentage change rates or differences to previous period, year

```
ts_pc(fdeaths)
```



ts_scale(): normalize mean and variance

```
ts_scale(fdeaths)
```



ts_index(): Index, based on levels

ts_compound(): Index, based on growth rates

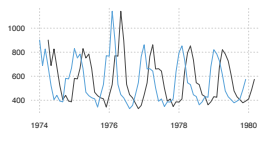
```
ts_index(fdeaths, base = 1976)
```



ts_seas(): seasonal adjustment using X-13

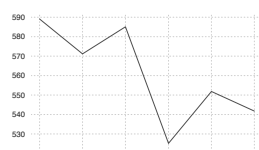
```
ts_seas(fdeaths)
```

SPAN AND FREQUENCY



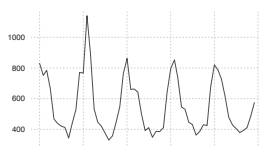
ts_lag(): Lag or lead of time series

```
ts_lag(fdeaths, 4)
```



ts_frequency(): convert to frequency

```
ts_frequency(fdeaths, "year")
```



ts_span(): filter time series for a time span.

```
ts_span(fdeaths, "1976-01-01")
ts_span(fdeaths, "-5 year")
```

Class Conversion

tsbox is built around a set of converters, which convert time series of the following **supported classes** to each other:

converter function	ts-boxable class
<code>ts_ts()</code>	ts, mts
<code>ts_data.frame(), ts_df()</code>	data.frame
<code>ts_data.table(), ts_dt()</code>	data.table
<code>ts_tbl()</code>	df_tbl, "tibble"
<code>ts_xts()</code>	xts
<code>ts_zoo()</code>	zoo
<code>ts_tibbltime()</code>	tibbltime
<code>ts_timeSeries()</code>	timeSeries
<code>ts_tsibble()</code>	tsibble
<code>ts_tslist()</code>	a list with ts objects

Time Series in data frames

LONG STRUCTURE

Default structure to store multiple time series in long data frames (or data tables, or tibbles)

```
ts_df(ts_c(fdeaths, mdeaths))
```

id	time	value
fdeaths	1974-01-01	901
fdeaths	1974-02-01	689
fdeaths	1974-03-01	827
...

AUTO-DETECT COLUMN NAMES

tsbox auto-detects a *value*-, a *time*- and zero, one or several *id*-columns. Alternatively, the *time*- and the *value*-column can be explicitly named **time** and **value**.

ts_default(): standardize column names in data frames

RESHAPE

ts_wide(): convert default long structure to wide

ts_long(): convert wide structure to default long

USE WITH PIPE

tsbox plays well with tibbles and with `%>%`, so it can be easily integrated into a dplyr/pipe workflow

```
library(dplyr)
ts_c(fdeaths, mdeaths) %>%
  ts_tbl() %>%
  ts_trend() %>%
  ts_pc()
```

pass return value as first argument to the next function