

The ARMAX Model

Full Name:

The **A**utoregressive **M**oving **A**verage **eX**ogenous Model

Mathematical Notation:

$$P_t = c + \beta X + \varphi_1 P_{t-1} + \theta_1 \varepsilon_{t-1} + \varepsilon_t$$

P_t, P_{t-1}

Values in the current period and 1 period ago respectively

$\varepsilon_t, \varepsilon_{t-1}$

Error terms for the same two periods

c

Baseline constant factor

φ_1

What part of the value last period is relevant in explaining the current one

θ_1

What part of the error last period is relevant in explaining the current value

X

Exogenous variable

β

Coefficient for the exogenous variable

Description:

The ARMAX is an extension of the ARIMA model, which incorporates other **exogenous** variables.

These variables can be pretty much anything that can have an affect on the values we are trying to estimate. The only requirement is that we have data available for every time period we are interested in. Thus, we often rely on other time series as the exogenous components in the regression.

These models are great, when a big part of the change period to period cannot be explained by past values and past errors alone, so including other relevant values might be of great help (like the prices for an index of a market of a neighbouring country).

The ARMAX Model

Implementation of the Simple Model in Python:

The library the
ARMA method
comes from

The method we
are importing

```
from statsmodels.tsa.arima_model import ARMA
```

```
model_ar_1_ma_1_X_spx = ARMA(df.returns[1:], exog = df.returns_spx[1:], order=(1,1))
```

↑
The variable storing the
model characteristics
that we will fit later

↑
The time series we
wish to analyse

↑
The exogenous
variable we are
adding to the
ARMA model

↑
The order of the model

*For an ARMAX(p,q) model,
simply change the order
from (1,1) to (p,q).