

1.3: Some Simple Time Series Models

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defn

A **time series model** for the observed data x_1, x_2, \dots, x_T , is a specification of the joint distributions of a sequence of random variables X_1, \dots, X_T , of which x_1, \dots, x_T are postulated to be a realization.

It describes **all** cumulative distribution functions (CDFs) you can think of:

$$F_{X_{i_1}, \dots, X_{i_n}}(x_{i_1}, \dots, x_{i_n}) = P(X_{i_1} \leq x_{i_1}, \dots, X_{i_n} \leq x_{i_n})$$

for any n , and $i_1 < i_2 < \dots < i_n$.

It completely defines all behavior.

Usually we just look at **second order properties** of a time series.

This means we look at the **first-order** means $E[X_t]$, and the **second-order** expected products (building blocks of covariances) $E[X_t X_{t+h}]$, for any $t \in T_0$, and $h = 1, 2, \dots$

Sometimes this is sufficient. Sometimes, for complicated time series, it is not. What follows are a few simple models where it is sufficient to just know the second-order properties. In other words, where knowing the means and variances fully specifies a model.

Example 1: IID Noise

X_1, \dots, X_T are **iid noise** if they are

① **independent:**

$$F_{X_1, \dots, X_T}(x_1, \dots, x_n) = F_{X_1}(x_1) \times \dots, F_{X_T}(x_T),$$

for all $\{x_i\}$ and

② **identical:**

$$F_{X_i}(x_i) = F(x_i)$$

for all i , and

③ **noise**

$$E[X_t] = 0$$

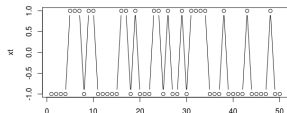
for all t

Example 1: IID Noise

IID noise has no trend, no seasonality, and no pattern in general. Sometimes they are assumed be all be **Normally distributed**. Or sometimes they are assumed to be iid **Bernoulli random variables** (a Binomial random variable when $n = 1$).

The outcomes of a Bernoulli rv are usually coded as 1 and 0. However, to get a mean of 0, we assume the outcomes are coded as 1 and -1 , and we assume $p = .5$.

```
plot(sample(x=c(1,-1), size=50,  
           replace=T, prob=c(.5,.5)), type = "b",  
      ylab = "xt")
```



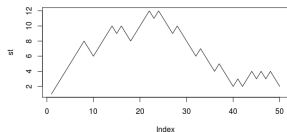
Example 2: Random Walk

A **(simple symmetric) random walk** is obtained by summing iid noise.
For each $t = 1, 2, \dots$

$$S_t = X_1 + X_2 + \dots + X_t = \sum_{i=1}^t X_i.$$

Usually it's convention to set $S_0 = 0$ with probability 1.

```
plot(cumsum(sample(x=c(1,-1), size=50, replace=T,  
  prob=c(.5,.5))), type = "l", ylab = "st")
```



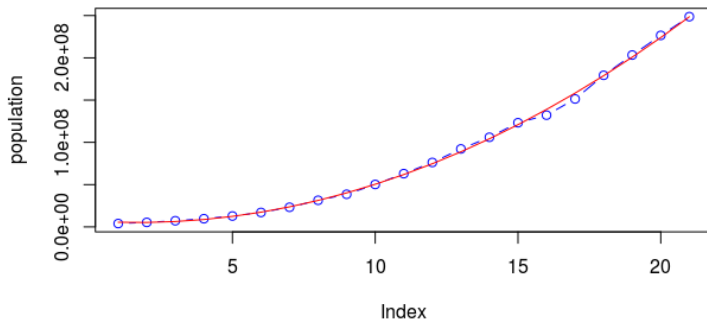
Example 3: Adding a Deterministic Trend

$$X_t = m_t + Y_t$$

- 1 X_t : object of interest
- 2 m_t : trend that we try to estimate
- 3 Y_t : iid noise

Example 3: Adding a Deterministic Trend

Example 1.3.4: A polynomial trend. Assume $m_t = a_0 + a_1 t + a_2 t^2$, for some unknown a_0 , a_1 and a_2 . See 1.3.R file.



Example 3: Adding a Deterministic Trend

```
> summary(fitMod)
```

Call:

```
lm(formula = population ~ t + I(t^2), data = popData)
```

Residuals:

Min	1Q	Median	3Q	Max
-6947521	-358167	436285	1481410	3391761

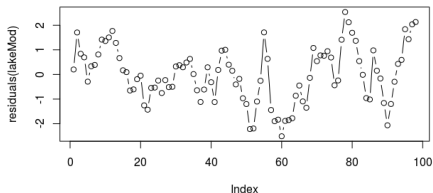
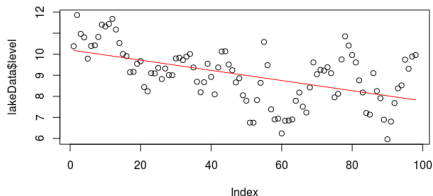
Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	6957920	1998526	3.482	0.00266	**
t	-2159870	418437	-5.162	6.55e-05	***
I(t^2)	650634	18472	35.223	< 2e-16	***

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Example 3: Adding a Deterministic Trend

Example 1.3.5: assume $m_t = a_0 + a_1 t$, for some unknown a_0 and a_1 . See 1.3.R file.



Example 4: Adding a Deterministic Trend

Harmonic Regression. Adding a seasonal trend.

$$X_t = s_t + Y_t$$

- ① X_t : object of interest
- ② s_t : periodic function that we try to estimate
- ③ Y_t : iid noise

Example 4: Adding a Deterministic Trend

Some background first:

- 1 **period:** units: time / 1 cycle
- 2 **frequency:** f , units: cycles / 1 time
- 3 frequency and period are reciprocals
- 4 Units of $2\pi f$ are radians / 1 time
- 5 Units of $2\pi ft$ are radians
- 6 Sometimes people write $\lambda = 2\pi f$ (angular frequency)
- 7 Angle-Sum Trig Identity: $\cos(a + b) = \cos(a)\cos(b) - \sin(a)\sin(b)$
- 8 Think unit circle: $\cos(0) = \cos(2\pi) = 1$, etc.

Example 4: Adding a Deterministic Trend

$$s_t = a_0 + \sum_{j=1}^k A_j \cos(2\pi f_j t + \phi_j) \quad \text{amps and phase}$$

$$= a_0 + \sum_{j=1}^k A_j \cos(2\pi f_j t) \cos(\phi_j) - A_j \sin(2\pi f_j t) \sin(\phi_j)$$

$$= a_0 + \sum_{j=1}^k [a_j \cos(2\pi f_j t) + b_j \sin(2\pi f_j t)] \quad \text{use this one}$$

Known frequencies: f_1, \dots, f_k

To be estimated: $a_0, a_1, \dots, a_k, b_1, \dots, b_k$

Example 4: Adding a Deterministic Trend

