# Time Series Cheatsheet

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Mean

$$E[X] = \int_{-\infty}^{\infty} x f(x) dx \text{ for } \int_{-\infty}^{\infty} |x| f(x) dx < \infty, \text{ undefined otherwise}$$

$$\mu_t = E[Y_t] \text{ for } t = 0, \pm 1, \pm 2, \dots$$

$$E[aX + bY + c] = aE[X] + bE[Y] + c$$

$$E[XY] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} xy f(x, y) dx dy$$

#### Variance

$$Var(X) \ge 0$$
 non-negative 
$$Var(a+bX) = b^2Var(X)$$
 
$$Var(X+Y) = Var(X) + Var(Y) \text{ for independent } X,Y$$
 
$$Var(X) = E[X^2] - \left(E[X]\right)^2$$

#### Covariance

$$Cov(a + bX, c + dY) = bdCov(X, Y)$$

$$Var(X + Y) = Var(X) + Var(Y) + 2Cov(X, Y)$$

$$Cov(X + Y, Z) = Cov(X, Z) + Cov(Y, Z)$$

$$Cov(X, X) = Var(X)$$

$$Cov(X, Y) = Cov(Y, X)$$

$$Cov(X, Y) = 0 \text{ for independent } X, Y$$

$$Cov(X, Y) = E[(X - E[X])(Y - E[Y])]$$

$$= E[(X - \mu_X)(Y - \mu_Y)]$$

$$= E[XY - XE[Y] - YE[X] + E[X]E[Y]]$$

$$= E[XY] - E[XE[Y]] - E[YE[X]] + E[E[X]E[Y]]$$

$$= E[XY] - E[X]E[Y]$$

#### Correlation

If  $X^*$  is a standardized X and  $Y^*$  is a standardized Y, then

$$\rho = E[X^*Y^*]$$

Else,

$$\rho = Corr(X, Y) = \frac{Corr(X, Y)}{\sqrt{Var(X)Var(Y)}}$$

$$\begin{aligned} -1 &\leq Corr(X,Y) \leq 1 \\ Corr(a+bX,c+dY) &= sign(bd)Corr(X,Y) \\ Corr(X,Y) &= \pm 1iff\exists a,b \text{ s.t. } P(Y=a+bX) = 1 \end{aligned}$$

**Events** 

$$e_i = \operatorname{event} i$$
  
 $Y_0 = e_0$   
 $Y_t = e_0 + e_1 + e_2 + \ldots + e_t$ 

#### Autocovariance

$$\gamma_{t,t} = Var(Y_t)$$

$$\gamma_{t,s} = \gamma_{s,t}$$

$$\left|\gamma_{t,s}\right| \le \sqrt{\gamma_{t,t}\gamma_{s,s}}$$

$$\gamma_{t,s} = Cov(Y_t, Y_s) \text{ for } t, s = 0, \pm 1, \pm 2, \dots$$

$$Cov(X, Y) = E\left[(X - \mu_X)(Y - \mu_Y)\right]$$

$$= E[XY] - E[X]E[Y]$$

$$Cov(Y_t, Y_s) = E\left[(Y_t - \mu_t)(Y_s - \mu_s)\right]$$

$$= E[Y_t Y_s] - \mu_t \mu_s$$

For stochastic process  $e_1, e_2, \ldots$  with mean 0, variance  $\sigma^2$ ,

$$\begin{split} &\gamma_{t,t} = Var(Y_t) \\ &= Var(\sum_{i=1}^t e_i) \\ &= \sum_{i=1}^t Var(e_i) \\ &= t\sigma^2 \\ &\gamma_{t,s} = Cov(Y_t, Y_s) \\ &= Cov(e_1 + e_2 + \ldots + e_t, e_1, e_2, \ldots + e_s) \\ &= Cov([\sum_{i=1}^n c_i Y_{t_i}, \sum_{j=1}^n d_j Y_{s_j}]) \\ &= \sum_{i=1}^m \sum_{j=1}^n c_i d_j Cov(Y_{t_i}, Y_{s_j}) \\ &= \sum_{i=1}^t \sum_{j=1}^s Cov(e_i, e_j) \text{ (when } i = j, \text{ you get variance, when } i \neq j \text{ you get 0)} \\ &\text{ for } 1 \leq t \leq s \\ &= t\sigma^2 \end{split}$$

If  $c_1, c_2, \ldots, c_m$  and  $d_1, d_2, \ldots, d_n$  are constants and  $t_1, t_2, \ldots, t_m$  and  $s_1, s_2, \ldots, s_n$  are time points, then

$$Cov\left[\sum_{i=1}^{m} c_{i}Y_{t_{i}}, \sum_{j=1}^{n} d_{j}Y_{s_{j}}\right] = \sum_{i=1}^{m} c_{i}d_{j}Cov(Y_{t_{i}}, Y_{s_{j}})$$

Special case

$$Var\left[\sum_{i=1}^{n} c_{i} Y_{t_{i}}\right] = \sum_{i=1}^{n} c_{i}^{2} Var(Y_{t_{i}}) + 2\sum_{i=2}^{n} \sum_{j=1}^{i-1} c_{i} c_{j} Cov(Y_{t_{i}}, Y_{t_{j}})$$

### Autocorrelation

$$\rho_{t,t} = 1$$

$$\rho_{t,s} = \rho_{s,t}$$

$$|\rho_{t,s}| \le 1$$

$$\rho_{t,s} = \frac{\gamma_{t,s}}{\sqrt{\gamma_{t,t}, \gamma_{s,s}}}$$

# **Strictly Stationary**

The joint distribution of  $Y_{t_1}, Y_{t_2}, \dots, Y_{t_n}$  is the same as  $Y_{t_1-k}, Y_{t_2-k}, \dots, Y_{t_n-k} \forall t_i, i=1,2,\dots,n$  and lag k. Also,  $Var(Y_t) = Var(Y_{t-k})$  (constant variance over time).

 $Cov(Y_{t-k}, Y_{s-k}) \forall t, s \text{ and } k.$  Putting k = s and k = t, we get

$$\begin{split} \gamma_{t,s} &= Cov(Y_{t-s}, Y_0) \\ &= Cov(Y_0, Y_{s-t}) \\ &= Cov(Y_0, Y_{|t-s|}) \\ &= \gamma_{0,|t-s|} \\ \gamma_k &= Cov(Y_t, Y_{t-k}) \\ \rho_k &= Corr(Y_t, Y_{t-k}) \\ \rho_k &= \frac{\gamma_k}{\gamma_0} \\ \gamma_0 &= Var(Y_t) \\ \gamma_k &= \gamma_{-k} \\ |\gamma_k| &\leq \gamma_0 \\ \rho_0 &= 1 \\ \rho_k &= \rho_{-k} \\ |\rho_k| &\leq 1 \end{split}$$

## Weakly Stationary

- 1. Mean function is **constant over time**
- 2.  $\gamma_{t,t-k} = \gamma_{0,k}$  for all time t and lag k