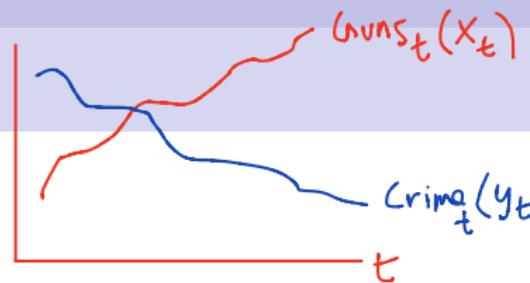


## Tutorial 4: Time Series Seasonality and Stationarity

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## Seasonal Adjustments



$\Rightarrow$  Regress  $\text{crime}_t$  on  $\text{gunst}_t$   
 $\Rightarrow \hat{\beta}_1 < 0$

$\hookrightarrow$  Could be case that  
 gun ownership does  
 not cause crime, but  
 both have a trend.

- Challenge: series tend to trend with seasons and time

$\hookrightarrow$  Need to isolate seasonality & trends  
 to avoid spurious regression results

- Include time trends in regression

$$\hookrightarrow y_t = \beta_0 + \beta_1 x_t + \underbrace{\beta_2 t}_{\text{linear trend}} + \varepsilon_t$$

- Include time indicator variables in regression

$$\hookrightarrow y_t = \beta_0 + \beta_1 x_t + \underbrace{\beta_t^*}_{\text{Seasonal fixed effects}} + \varepsilon_t, \quad t = \text{monthly level, } t^* = \text{Quarter of year}$$

→ If stationarity & WD hold  $\Rightarrow$  relax OLS assumption

## Stationary and Weakly Dependant Time Series

Strong Stationarity

- Stationary: joint distribution of stochastic process is stable over time  $F_x(x_{t_1}, x_{t_2}, \dots, x_{t_n}) = F_x(x_{t_1+h}, x_{t_2+h}, \dots, x_{t_n+h})$

$\Rightarrow$  same statistical charac. across equal size windows  
↳ Constant mean, variance, skewness, etc.

Weak Stationarity

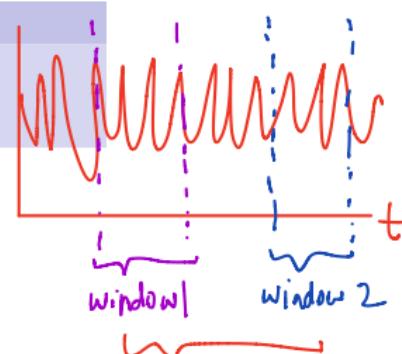
- Covariance stationary: finite mean, finite variance, and  $\text{Cov}(x_t, x_{t+h})$  only depends on  $h$

$$\hookrightarrow E(x_t) = \mu \text{ and } \text{Cov}(x_t, x_{t+h}) = g(h)$$

- Weakly dependant:  $\text{Cov}(x_t, x_{t+h}) \rightarrow 0$  has  $h \rightarrow \infty$

↳ Replacing "independent" assumption from

Cross-sectional data  $\Rightarrow$  Enables for  
the usage of LLN & CLT for inference



Stationarity implies  
both windows of  
equal length realized  
from same distn.

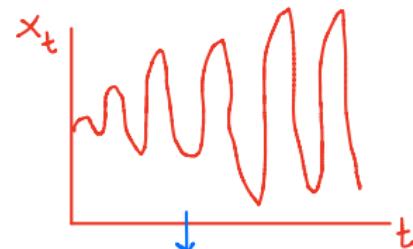
## Autoregressive Model and Examples

→ If  $\rho=1 \Rightarrow V(y_t) = \infty \& \text{corr}(y_t, y_{t+h}) = 1 \Rightarrow \text{not } \textcircled{S} \Rightarrow \text{not WD}$   
 ↳ similar argument for  $\rho=-1$

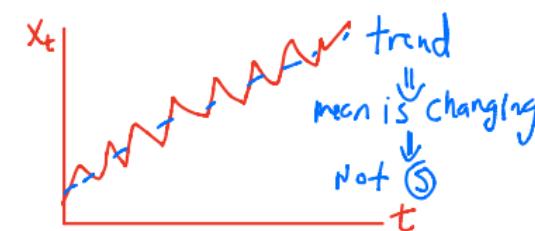
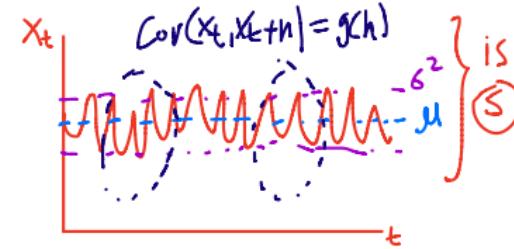
- AR(1) model:  $y_t = \rho y_{t-1} + u_t$ , AR(2):  $y_t = \rho_1 y_{t-1} + \rho_2 y_{t-2} + u_t$ ,  $u_t \sim N(0, \sigma^2_u)$

↳ Show  $E(y_t) = 0$ ,  $V(y_t) = \frac{\sigma^2_u}{1-\rho^2}$ ,  $\text{corr}(y_t, y_{t+h}) = \rho^h \Rightarrow$  stationarity  $\Rightarrow$  WD requires  $\rho \in (-1, 1)$

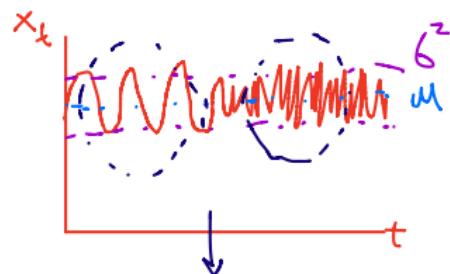
- Which one of the following time series is stationary?



mean constant but  
 var  $\uparrow \Rightarrow$  not  $\textcircled{S}$



trend  
 mean is changing  
 not  $\textcircled{S}$



Covariance structure changes  
 with time  $\Rightarrow$  not  $\textcircled{S}$

## Seasonal adjustment practice

### Chapter 10 Problem 5

Suppose you have quarterly data on new housing starts, interest rates, and real per capita income. Specify a model for housing starts that accounts for possible trends and seasonality in the variables.

$$HStart_t = \beta_0 + \beta_1 \text{int}_t + \beta_2 \text{rpcit}_t + \underbrace{\beta_3 Q2_t + \beta_4 Q3_t + \beta_5 Q4_t}_{\text{seasonality}} + \underbrace{\beta_6 t}_{\text{trend}} + \epsilon_t$$

# Seasonal adjustment practice

## Chapter 11 Problem 2

$e_t \sim \text{iid}(0, 1) \Rightarrow e_t \text{ stationary}$

Let  $\{e_t : t = -1, 0, 1, \dots\}$  be a sequence of independent, identically distributed random variables with mean zero and variance one. Define a stochastic process by

$$x_t = e_t - (1/2)e_{t-1} + (1/2)e_{t-2}, t = 1, 2, \dots$$

- ① Find  $E(x_t)$  and  $\text{Var}(x_t)$ . Do either of these depend on  $t$ ?

$$\begin{aligned} E(x_t) &= E(e_t) - \frac{1}{2}E(e_{t-1}) + \frac{1}{2}E(e_{t-2}) = 0 \\ V(x_t) &= V(e_t) + \frac{1}{4}V(e_{t-1}) + \frac{1}{4}V(e_{t-2}) = 1 + \frac{1}{4} + \frac{1}{4} = \frac{3}{2} \end{aligned} \quad \left. \begin{array}{l} \text{not depend} \\ \text{on } t \end{array} \right\}$$

- ② What is  $\text{Corr}(x_t, x_{t+h})$  for  $h > 2$ ?

$$\text{Corr}(x_t, x_{t+h}) = \frac{\text{Cov}(x_t, x_{t+h})}{\sqrt{V(x_t)V(x_{t+h})}}, \text{Cov}(\underbrace{e_t - \frac{1}{2}e_{t-1} + \frac{1}{2}e_{t-2}}_{x_t}, \underbrace{e_{t+h} - \frac{1}{2}e_{t+h-1} + \frac{1}{2}e_{t+h-2}}_{x_{t+h}}) = 0 \quad h > 2$$

- ③ Is  $x_t$  an asymptotically uncorrelated process?

As  $h \rightarrow \infty \Rightarrow \text{Corr}(x_t, x_{t+h}) \rightarrow 0 \Rightarrow \text{Yes}$