

Analysis of Time Series, L9

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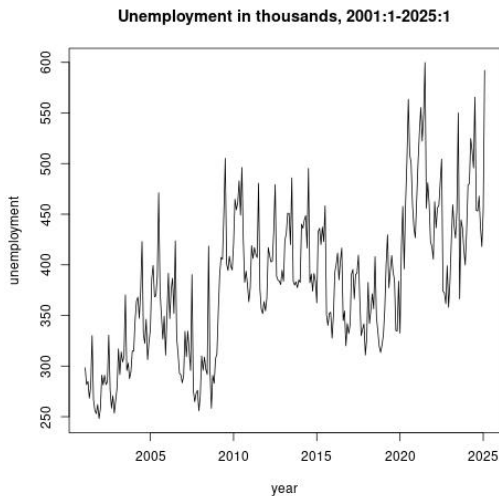
9 april 2025

Today

3.9: Multiplicative Seasonal ARIMA (SARIMA) Models

- SARIMA models
- Empirical examples
- Menti

SARIMA models



SARIMA models

- Recall: $\text{ARMA}(p, q)$

$$x_t = \phi_1 x_{t-1} + \dots + \phi_p x_{t-p} + w_t + \theta_1 w_{t-1} + \dots + \theta_q w_{t-q}$$

- or equivalently $\phi(B)x_t = \theta(B)w_t$ where

$$\phi(B) = 1 - \phi_1 B - \dots - \phi_p B^p,$$

$$\theta(B) = 1 + \theta_1 B + \dots + \theta_q B^q.$$

- Pure seasonal ARMA model*, $\text{ARMA}(P, Q)_s$

$$x_t = \Phi_1 x_{t-s} + \dots + \Phi_P x_{t-Ps} + w_t + \Theta_1 w_{t-s} + \dots + \Theta_Q w_{t-Qs}$$

- or equivalently $\Phi_P(B^s)x_t = \Theta_Q(B^s)w_t$ where

$$\Phi_P(B^s) = 1 - \Phi_1 B^s - \dots - \Phi_P B^{Ps},$$

$$\Theta_Q(B^s) = 1 + \Theta_1 B^s + \dots + \Theta_Q B^{Qs}.$$

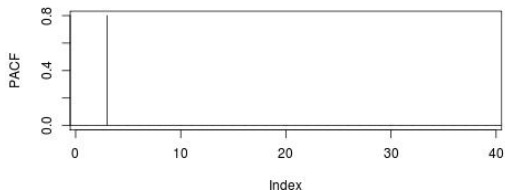
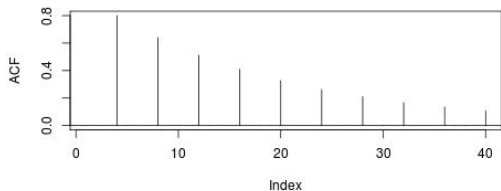
SARIMA models

Calculate the ACF of

- 1 $\text{ARMA}(0, 1)_4$
- 2 $\text{ARMA}(1, 0)_4$

SARIMA models

ACF and PACF for $x_t = 0.8x_{t-4} + w_t$



SARIMA models

- Pure seasonal ARMA(P, Q)_s model

$$\Phi_P(B^s)x_t = \Theta_Q(B^s)w_t$$

where

$$\Phi_P(B^s) = 1 - \Phi_1 B^s - \dots - \Phi_p B^{Ps},$$

$$\Theta_Q(B^s) = 1 + \Theta_1 B^s + \dots + \Theta_Q B^{Qs}.$$

- Multiplicative (mixed) ARMA(p, q) \times (P, Q)_s model

$$\Phi_P(B^s)\phi(B)x_t = \Theta_Q(B^s)\theta(B)w_t$$

where in addition

$$\phi(B) = 1 - \phi_1 B - \dots - \phi_p B^p,$$

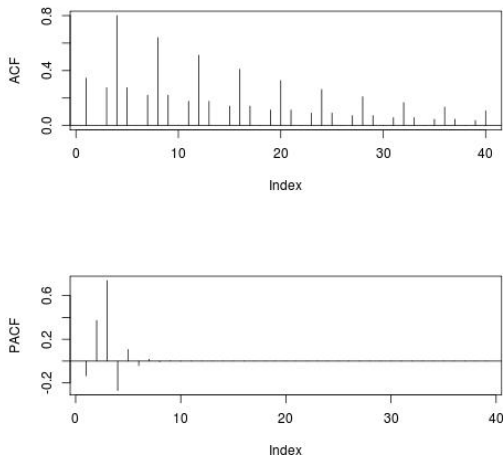
$$\theta(B) = 1 + \theta_1 B + \dots + \theta_q B^q.$$

SARIMA models

- 1 Write down the $\text{ARMA}(0, 1) \times (1, 0)_4$ model explicitly.
- 2 Derive its ACF.

SARIMA models

ACF and PACF for $x_t = 0.8x_{t-4} + w_t + 0.4w_{t-1}$



SARIMA models

Seasonal difference

$$\nabla_s x_t = (1 - B^s)x_t = x_t - x_{t-s}$$

Definition (3.12)

The multiplicative *seasonal autoregressive integrated moving average* (SARIMA) model is given by

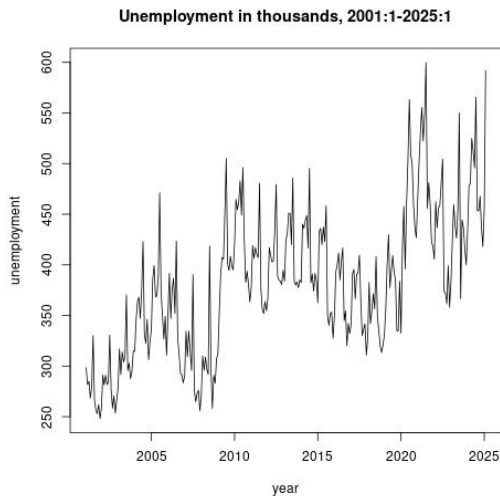
$$\Phi_P(B^s)\phi(B)\nabla_s^D\nabla^d x_t = \delta + \Theta_Q(B^s)\theta(B)w_t,$$

where w_t is white noise.

It is denoted by $\text{ARIMA}(p, d, q) \times (P, D, Q)_s$.

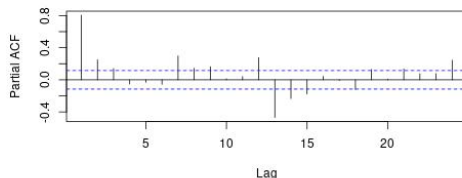
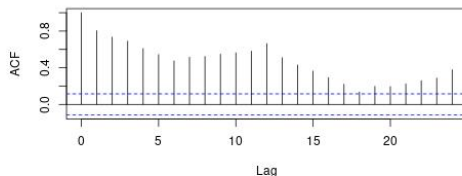
Write down the $\text{ARIMA}(0, 1, 1) \times (1, 1, 0)_4$ model without constant explicitly.

Empirical examples



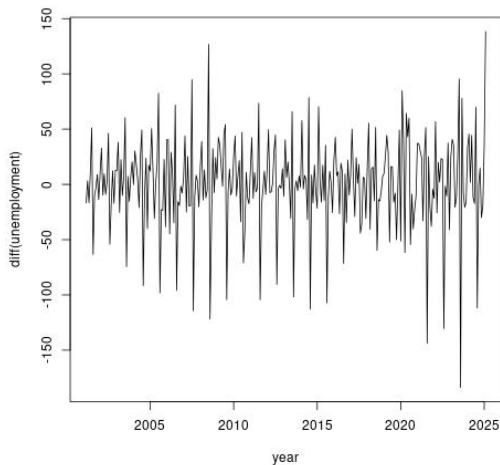
Empirical examples

Unemployment, ACF and PACF. Slowly decaying ACF.
Try to take a difference!



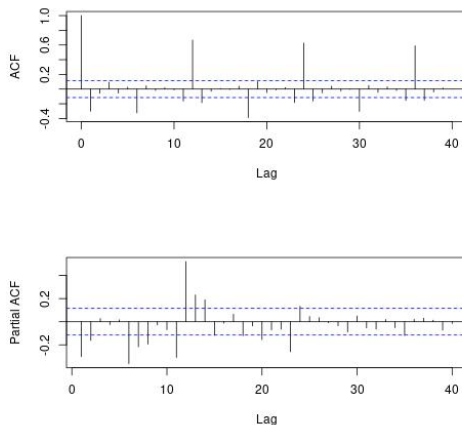
Empirical examples

Plot of the differenced series:



Empirical examples

Differenced series, ACF and PACF. Slowly decaying ACF at lags 12, 24, 36,... (R: `acf(diff(u),40)` etc.) Try a seasonal difference.



"Empty" seasonally differenced model:

```
> m=arima(u,order=c(0,1,0),seasonal=list(order=c(0,1,0),period=12));m
```

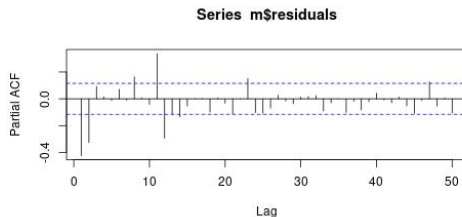
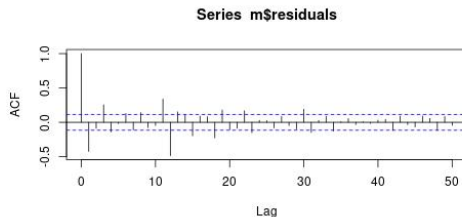
Call:

```
arima(x = u, order = c(0, 1, 0), seasonal = list(order = c(0, 1, 0), period = 12))
```

σ^2 estimated as 1054: log likelihood = -1352.2, aic = 2706.39

ACF and PACF for residuals: Seasonal MA(1) behavior?

```
> acf(m$residuals,50)
> pacf(m$residuals,50)
```



Try $\text{ARIMA}(0, 1, 0) \times (0, 1, 1)_{12}$ (smallest AIC of purely seasonal models).

```
> m=arima(u,order=c(0,1,0),seasonal=list(order=c(0,1,1),period=12));m
```

Call:

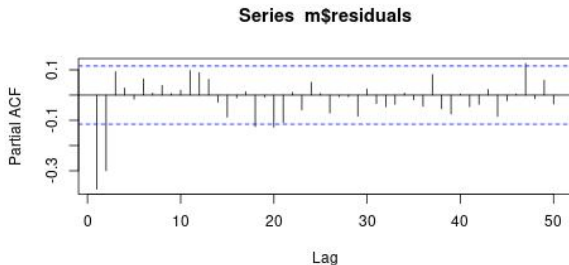
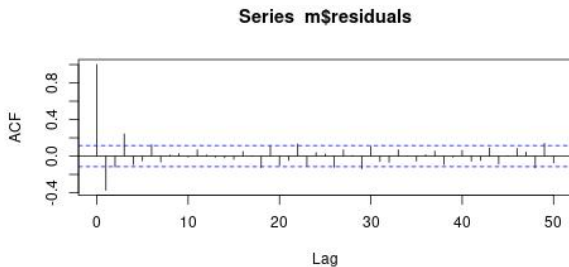
```
arima(x = u, order = c(0, 1, 0), seasonal = list(order = c(0, 1, 1), period = 12))
```

Coefficients:

```
      sma1  
      -0.8687  
s.e.      0.0562
```

```
sigma^2 estimated as 626.1:  log likelihood = -1288.7,  aic = 2581.4
```

ACF and PACF for residuals: AR(2) behavior?



Try $\text{ARIMA}(2, 1, 0) \times (1, 1, 1)_{12}$. (Smaller AIC than the 'closest' models.)

```
> m=arima(u,order=c(2,1,0),seasonal=list(order=c(1,1,1),period=12));m
```

Call:

```
arima(x = u, order = c(2, 1, 0), seasonal = list(order = c(1, 1, 1), period = 12))
```

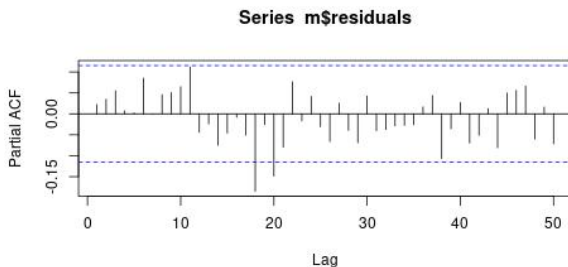
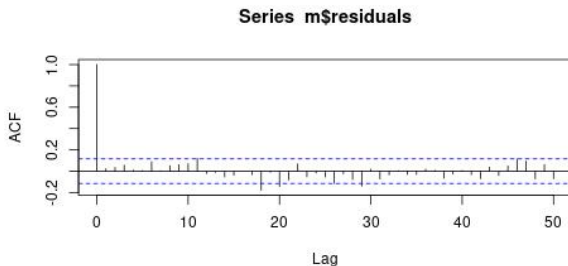
Coefficients:

| | ar1 | ar2 | sar1 | sma1 |
|------|---------|---------|--------|---------|
| | -0.5384 | -0.3427 | 0.1075 | -0.8991 |
| s.e. | 0.0592 | 0.0587 | 0.0748 | 0.0574 |

sigma^2 estimated as 471.3: log likelihood = -1250.03, aic = 2510.07

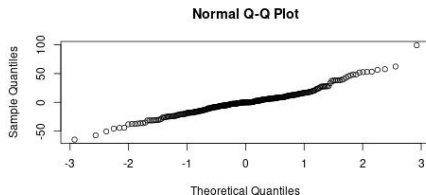
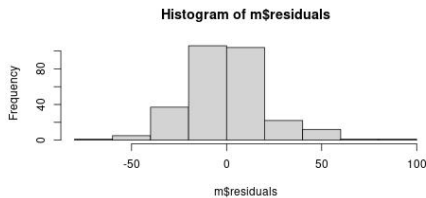
Observe: the sar1 coefficient is not significant.

ACF and PACF for residuals: White noise behavior?



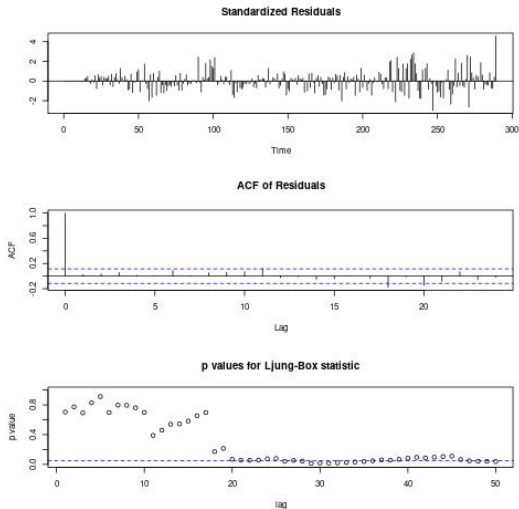
Histogram and qqplot for residuals:

```
> par(mfrow=c(2,1))  
> hist(m$residuals)  
> qqnorm(m$residuals)
```



Diagnostics for residuals: Too small P values!

```
> tsdiag(m,50)
```



Instead, try $\text{ARIMA}(3, 1, 0) \times (0, 1, 1)_{12}$. (A little bit higher AIC, but maybe better diagnostics?)

```
> m=arima(u,order=c(3,1,0),seasonal=list(order=c(0,1,1),period=12));m
```

Call:

```
arima(x = u, order = c(3, 1, 0), seasonal = list(order = c(0, 1, 1), period = 12))
```

Coefficients:

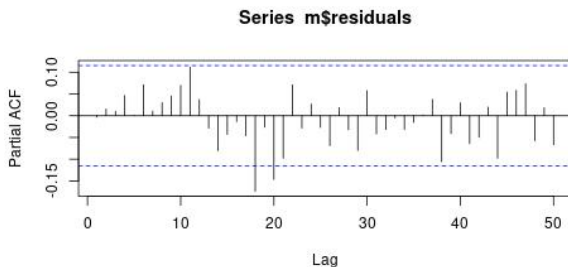
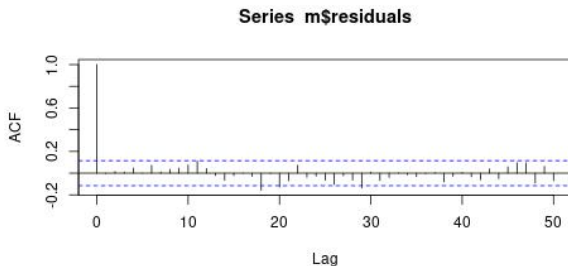
| | ar1 | ar2 | ar3 | sma1 |
|------|---------|---------|--------|---------|
| | -0.5065 | -0.3016 | 0.0704 | -0.8534 |
| s.e. | 0.0625 | 0.0679 | 0.0624 | 0.0532 |

sigma^2 estimated as 475.8: log likelihood = -1250.42, aic = 2510.85

Observe that the ar3 coefficient is not significant.

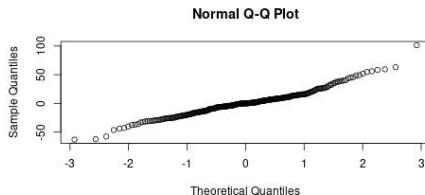
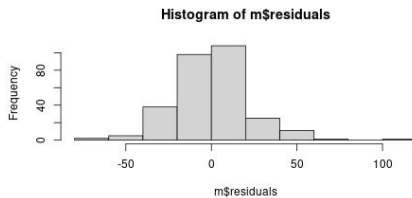
(But removing it gave worse Ljung-Box P values than for this model.)

ACF and PACF for residuals: White noise behavior?



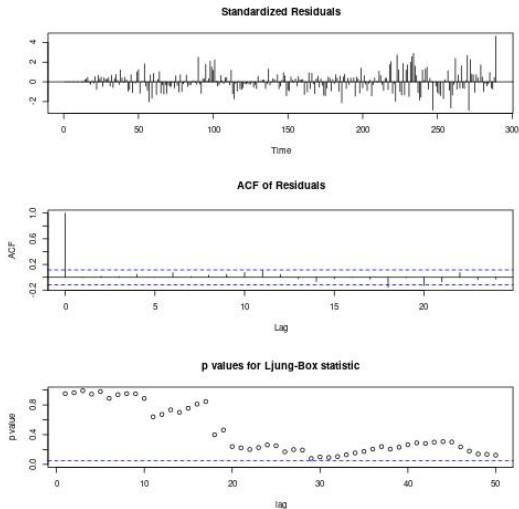
Histogram and qqplot for residuals:

```
> par(mfrow=c(2,1))  
> hist(m$residuals)  
> qqnorm(m$residuals)
```

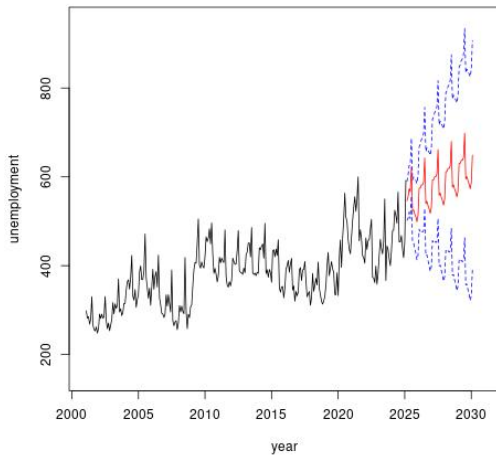


Diagnostics for residuals: Better P values!

```
> tsdiag(m,50)
```



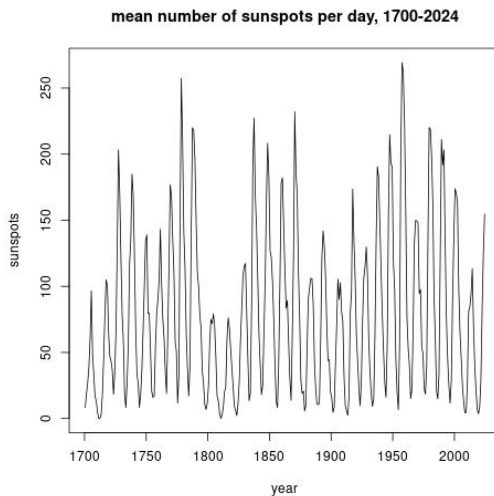
Plot of forecasts (5 years ahead):



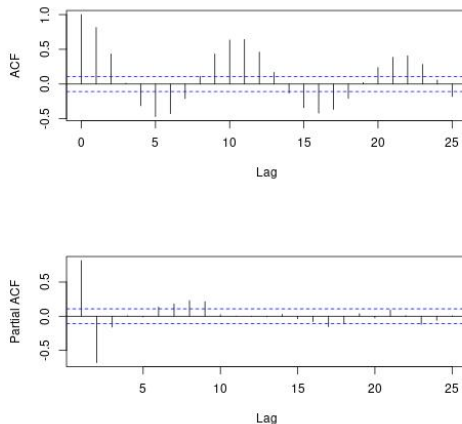
R code for this plot:

```
> fore=predict(m,n.ahead=60)
> plot(seq(2001+1/12,2025+1/12,1/12),u,type='l',xlab='year',
xlim=c(2001,2031),ylim=c(150,950),ylab='unemployment')
> lines(seq(2025+2/12,2030+1/12,1/12),fore$pred,col='red')
> lines(seq(2025+2/12,2030+1/12,1/12),fore$pred+2*fore$se,col='blue',
lty='dashed')
> lines(seq(2025+2/12,2030+1/12,1/12),fore$pred-2*fore$se,col='blue',
lty='dashed')
```

Empirical examples



Empirical examples



10-11 year season. AR(3) structure of PACF?

Try $\text{ARIMA}(3, 0, 0) \times (0, 0, 1)_{10}$ (smallest AIC among similar models):

```
> m=arima(s,order=c(3,0,0),seasonal=list(order=c(0,0,1),period=10));m
```

Call:

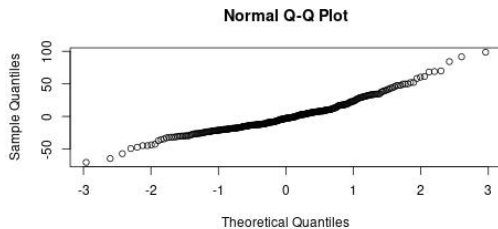
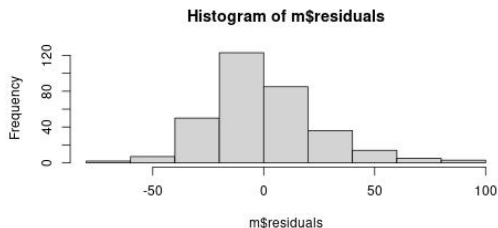
```
arima(x = s, order = c(3, 0, 0), seasonal = list(order = c(0, 0, 1), period = 10))
```

Coefficients:

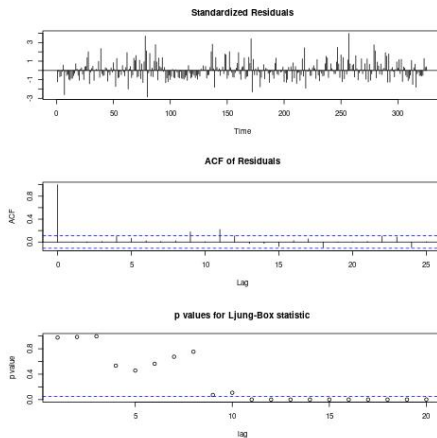
| | ar1 | ar2 | ar3 | sma1 | intercept |
|------|--------|---------|---------|--------|-----------|
| | 1.2052 | -0.4028 | -0.1670 | 0.2309 | 78.5312 |
| s.e. | 0.0587 | 0.0866 | 0.0548 | 0.0581 | 4.6094 |

σ^2 estimated as 611.7: log likelihood = -1505.21, aic = 3022.41

Empirical examples



Empirical examples



Small P values for Ljung-Box! Problem: the season is not exactly 10 years.

News of today

- Seasonal differencing
- SARIMA models