

Applied Time Series Analysis

SS 2013 – Week 08

Marcel Dettling

Institute for Data Analysis and Process Design

Zurich University of Applied Sciences

marcel.dettling@zhaw.ch

<http://stat.ethz.ch/~dettling>

ETH Zürich, April 15, 2013

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Non-Stationary Models: ARIMA and SARIMA

Why?

We have seen that many time series we encounter in practice show trends and/or seasonality. While we could decompose them and model the stationary part, it might also be attractive to directly model a non-stationary series.

How does it work?

There is a mechanism, "the integration" or "the seasonal integration" which takes care of the deterministic features, while the remainder is modeled using an ARMA(p,q).

There are some peculiarities!

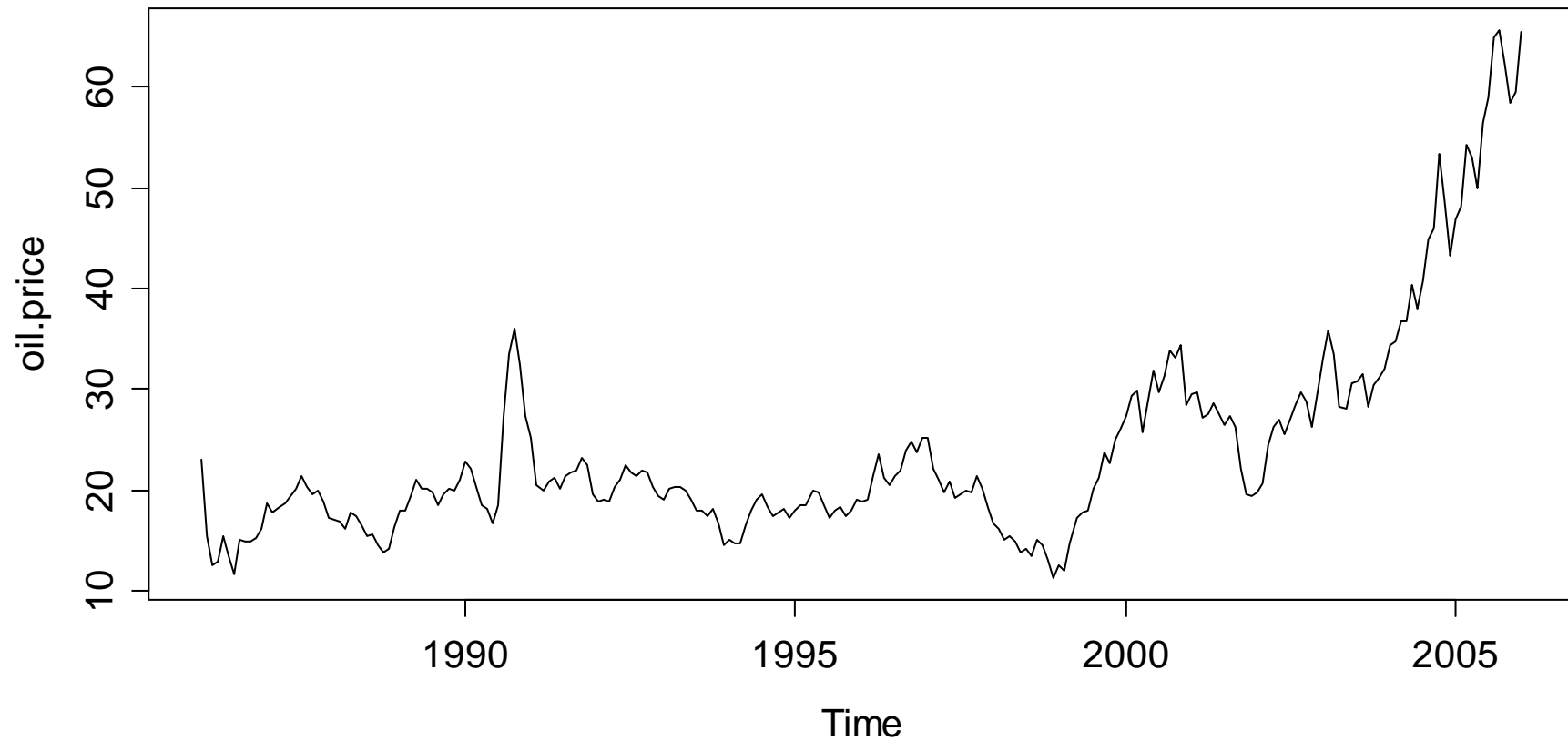
→ **see blackboard!**

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Example: Monthly Oil Prices

Monthly Price for a Barrel of Crude Oil

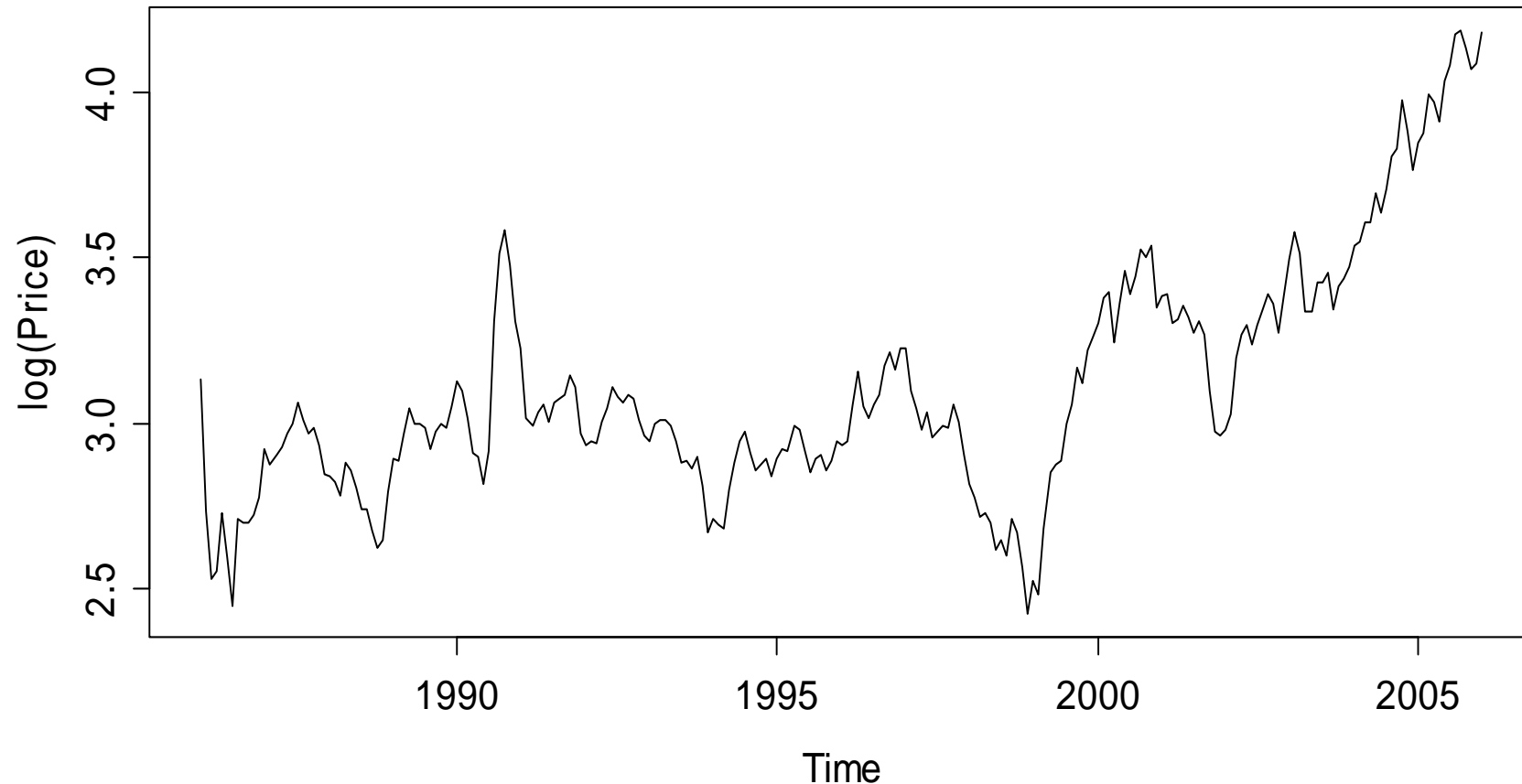


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Taking the Logarithm is Key

Logged Monthly Price for a Crude Oil Barrel

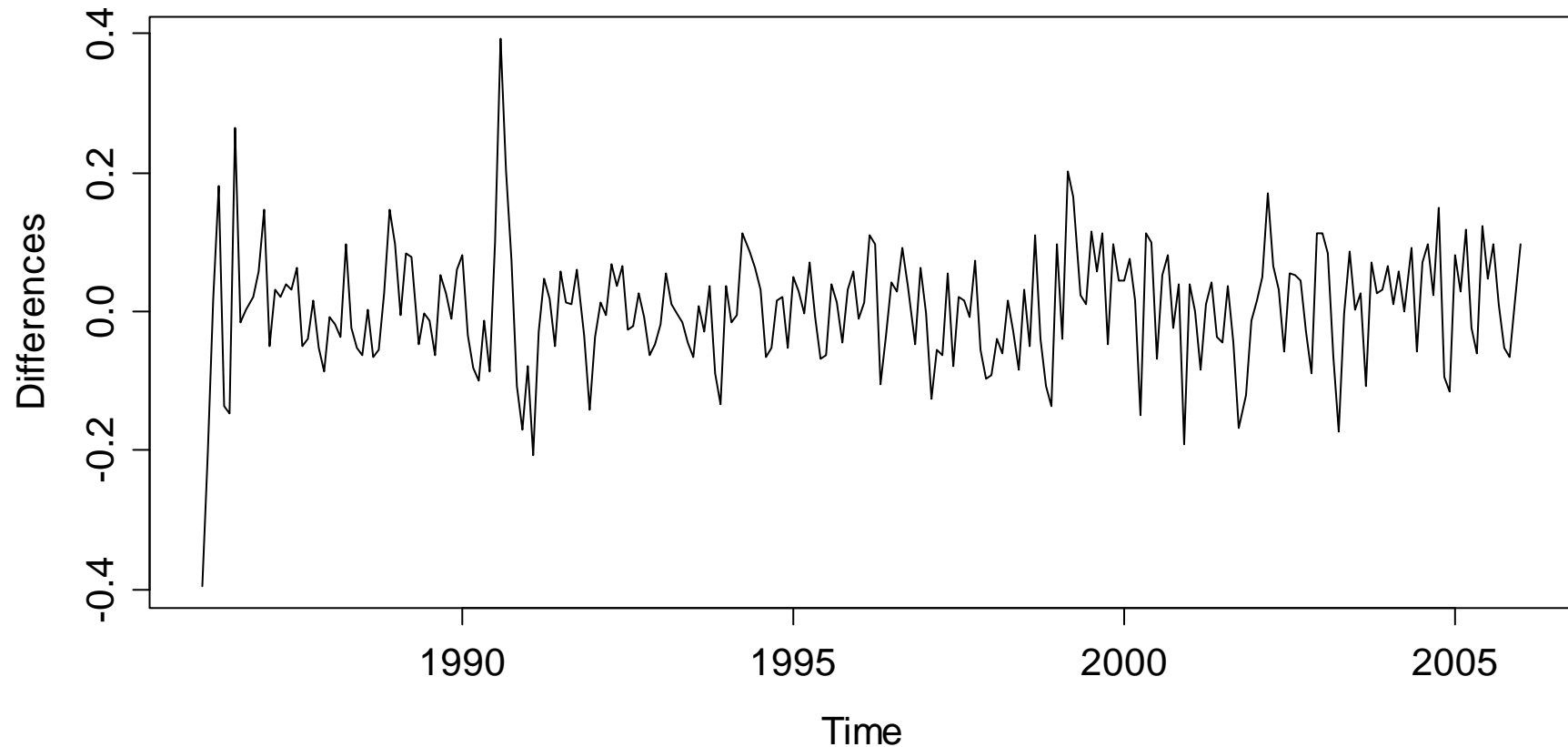


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Differencing Yields a Stationary Series

Differences of Logged Monthly Crude Oil Prices



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ARIMA(p,d,q)-Models

Idea: Fit an ARMA(p,q) to a time series where the dth order difference with lag 1 was taken before.

Example: If $Y_t = X_t - X_{t-1} = (1-B)X_t \sim ARMA(p, q)$,
then $X_t \sim ARIMA(p, 1, q)$

Notation: With backshift-operator B()

$$\Phi(B)(1-B)^d X_t = \Theta(B)E_t$$

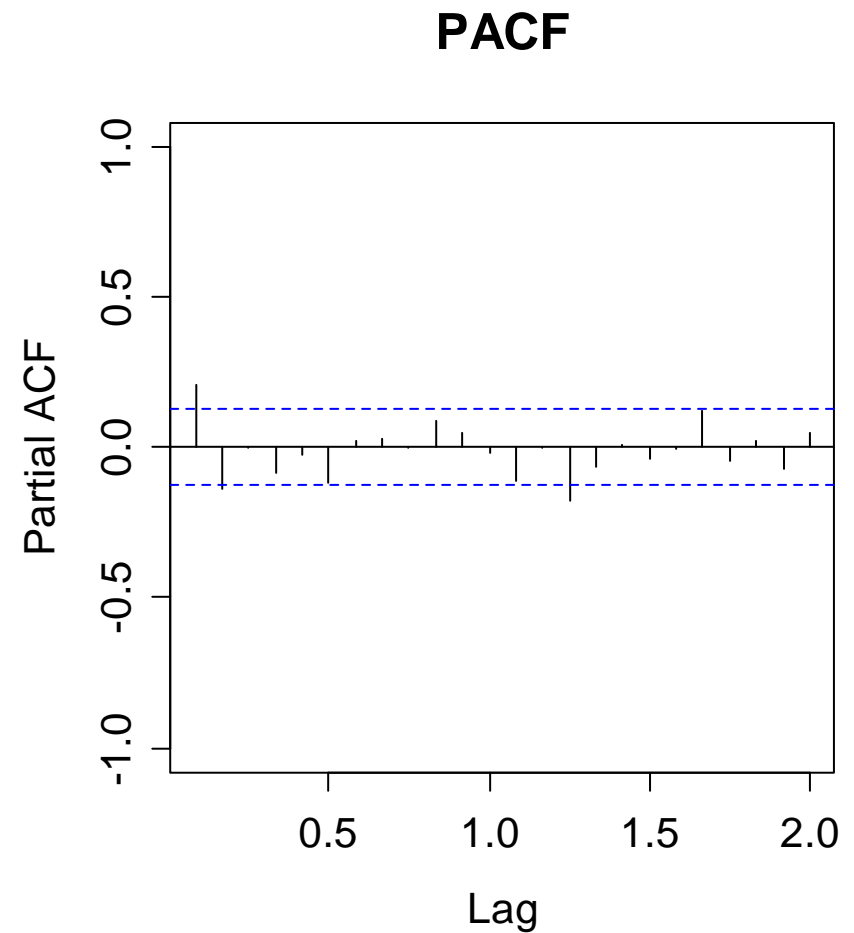
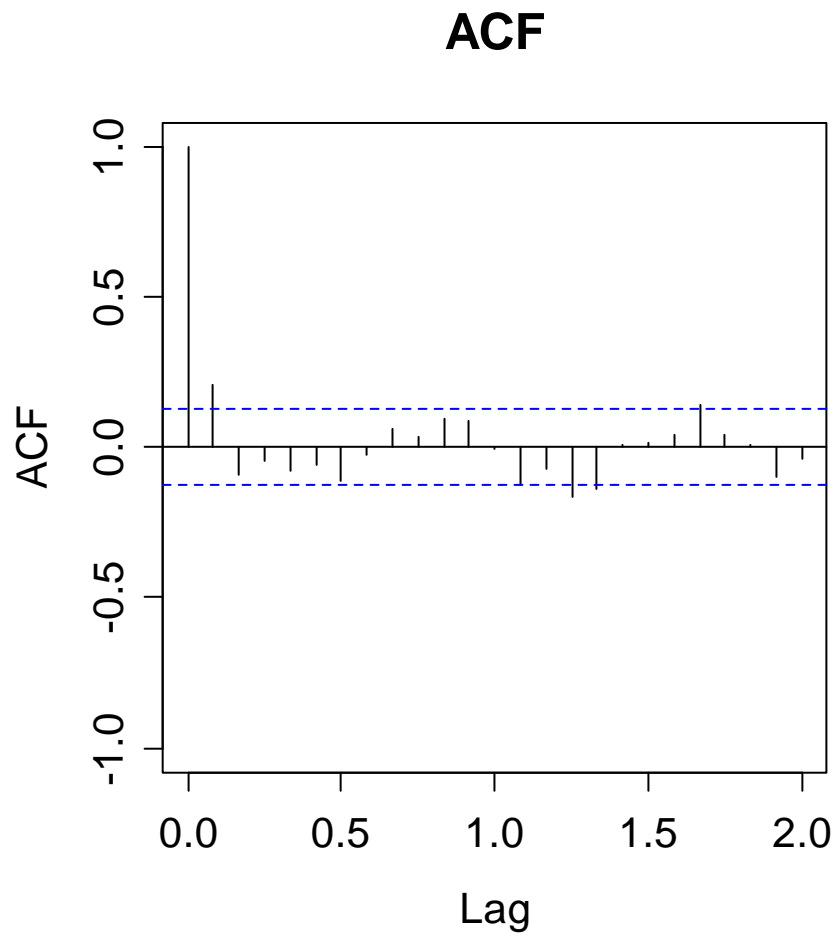
Stationarity: ARIMA-models are usually non-stationary!

Advantage: it's easier to forecast in R!

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ACF/PACF of the Differenced Series



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Fitting an ARIMA in R

We start by fitting an ARIMA(1,1,2) to the oil series:

```
> arima(lop, order=c(1,1,2))
```

Call:

```
arima(x = lop, order = c(1, 1, 2))
```

Coefficients:

| | ar1 | ma1 | ma2 |
|------|--------|---------|---------|
| | 0.8429 | -0.5730 | -0.3104 |
| s.e. | 0.1548 | 0.1594 | 0.0675 |

```
sigma^2 = 0.0066: ll = 261.88, aic = -515.75
```


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Alternative Fitting

Instead of fitting an ARIMA(1,1,2) to the logged oil series, we can also take the differenced log-oil series and fit an ARMA(1,2) to it.

IMPORTANT:

In this case, we have to do fitting without including an intercept (why?), thus:

```
> arima(diff(log(oil.price)), order=c(1,0,2),  
        include.mean=FALSE)
```

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Meaning of the Model / Recipe

We can rewrite the ARIMA(1,1,2) model as an ARMA(2,2),
see blackboard...

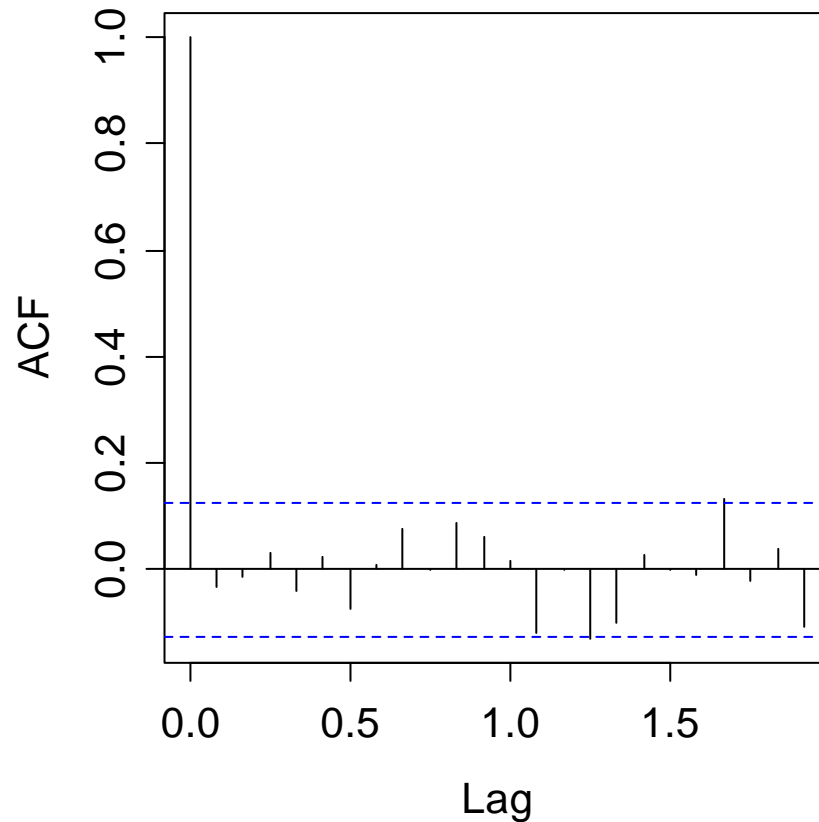
Some guidelines on how to fit ARIMA models to observed time series can also be found **on the blackboard...**

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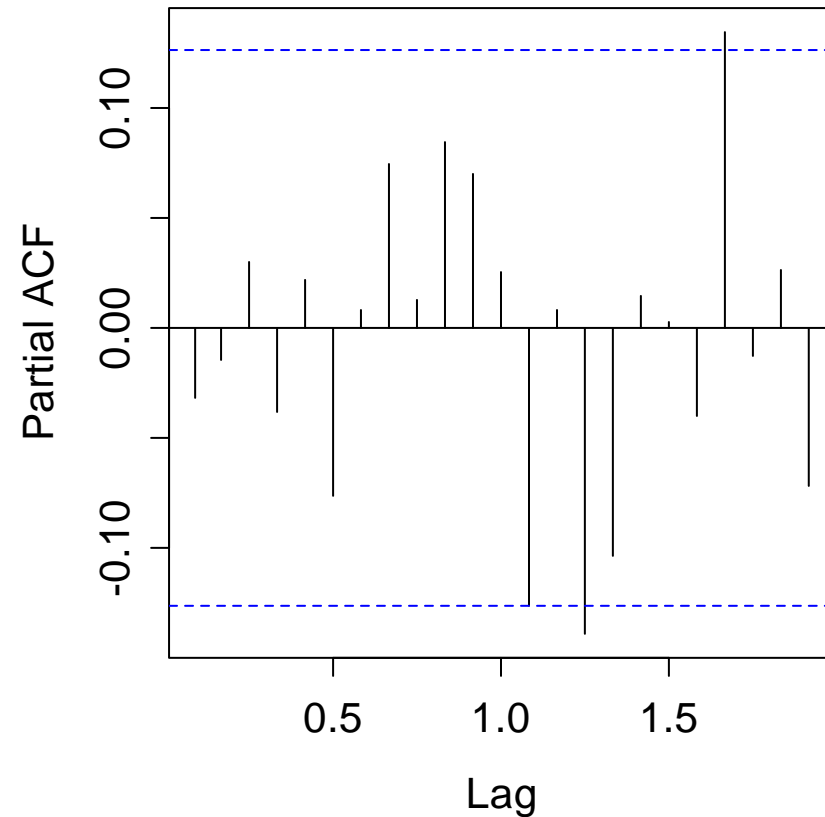
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Residual Analysis of the ARIMA(1,1,2)

ACF



PACF



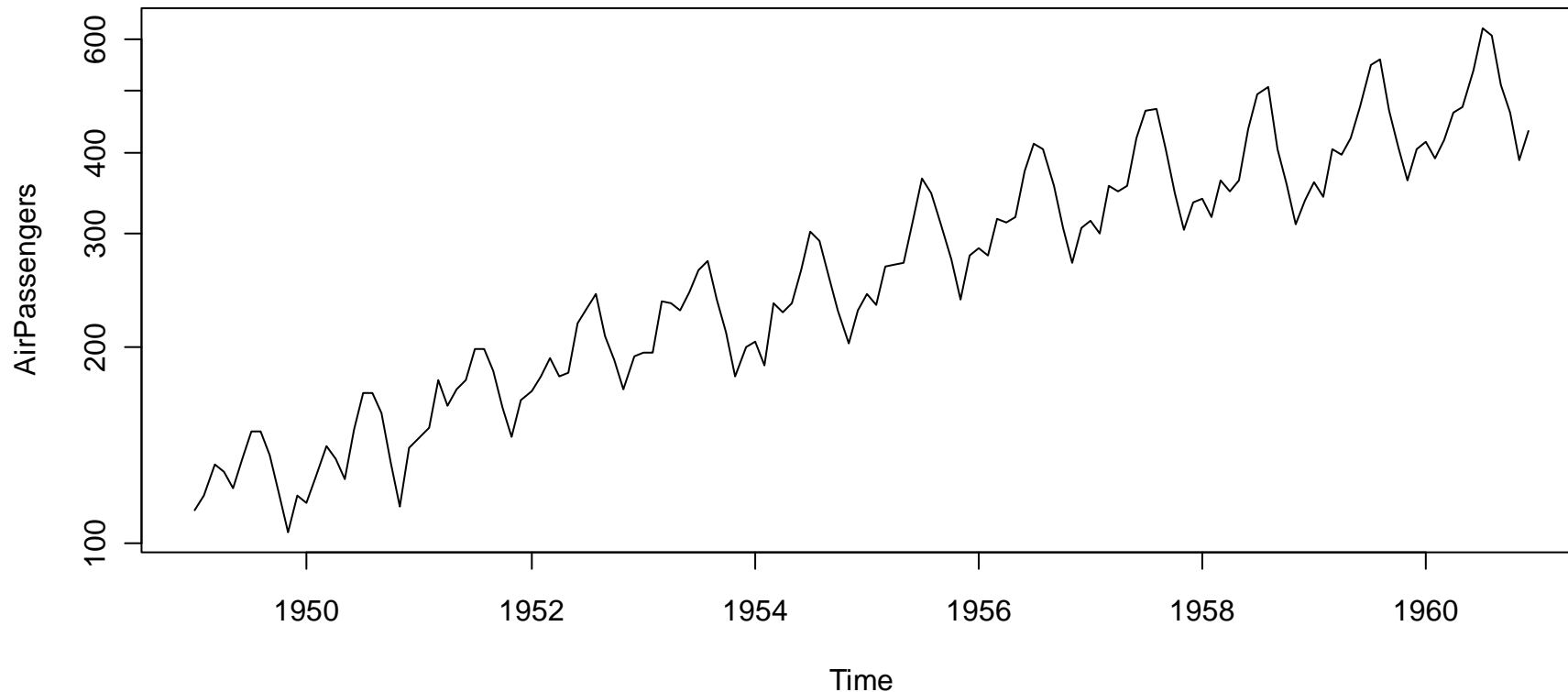
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$SARIMA(p,d,q)(P,D,Q)^s$

= a.k.a. Airline Model. We are looking at the log-trsf. airline data

Log-Transformed Airline Data

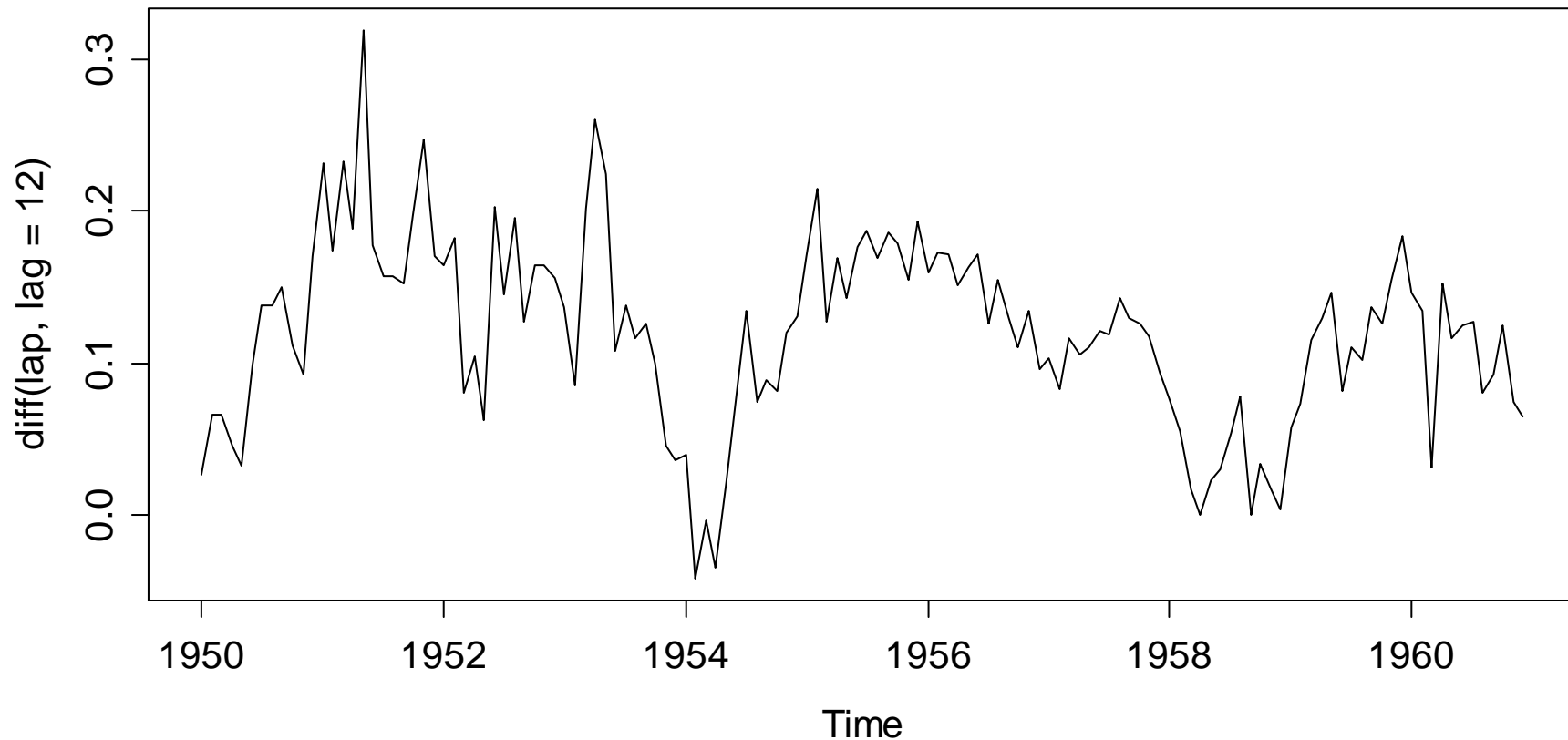


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Seasonal Differencing Helps...

Seasonally Differenced Airline Passenger Series

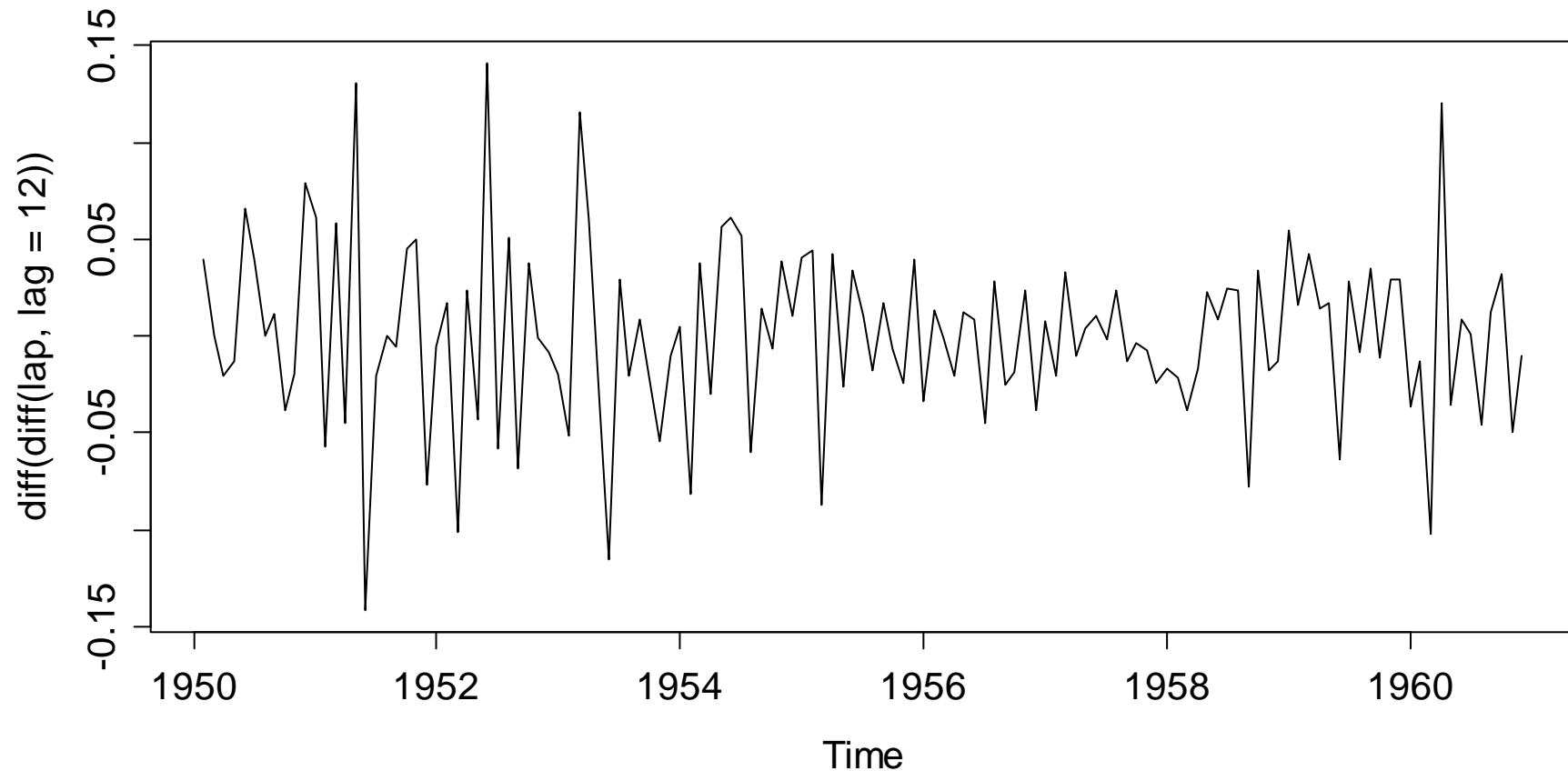


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... But More Is Needed!

Differenced Seasonally Differenced Airline Passenger Series



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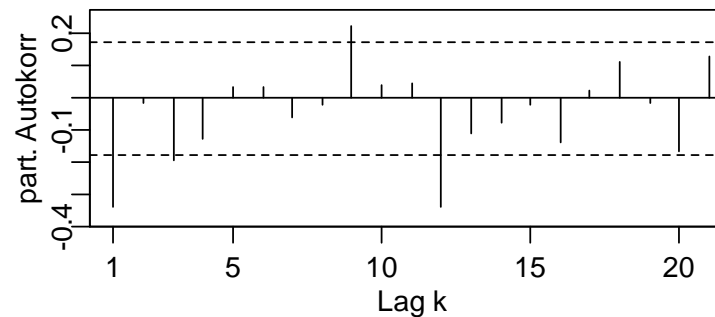
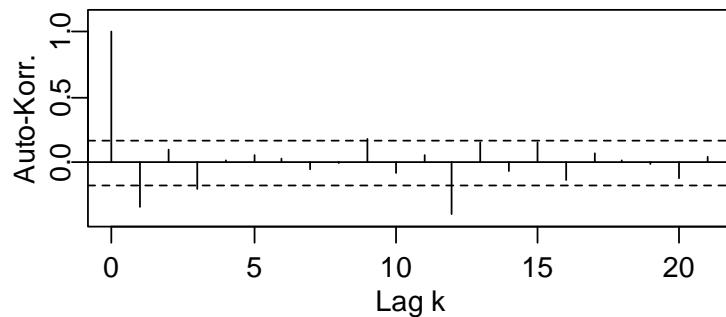
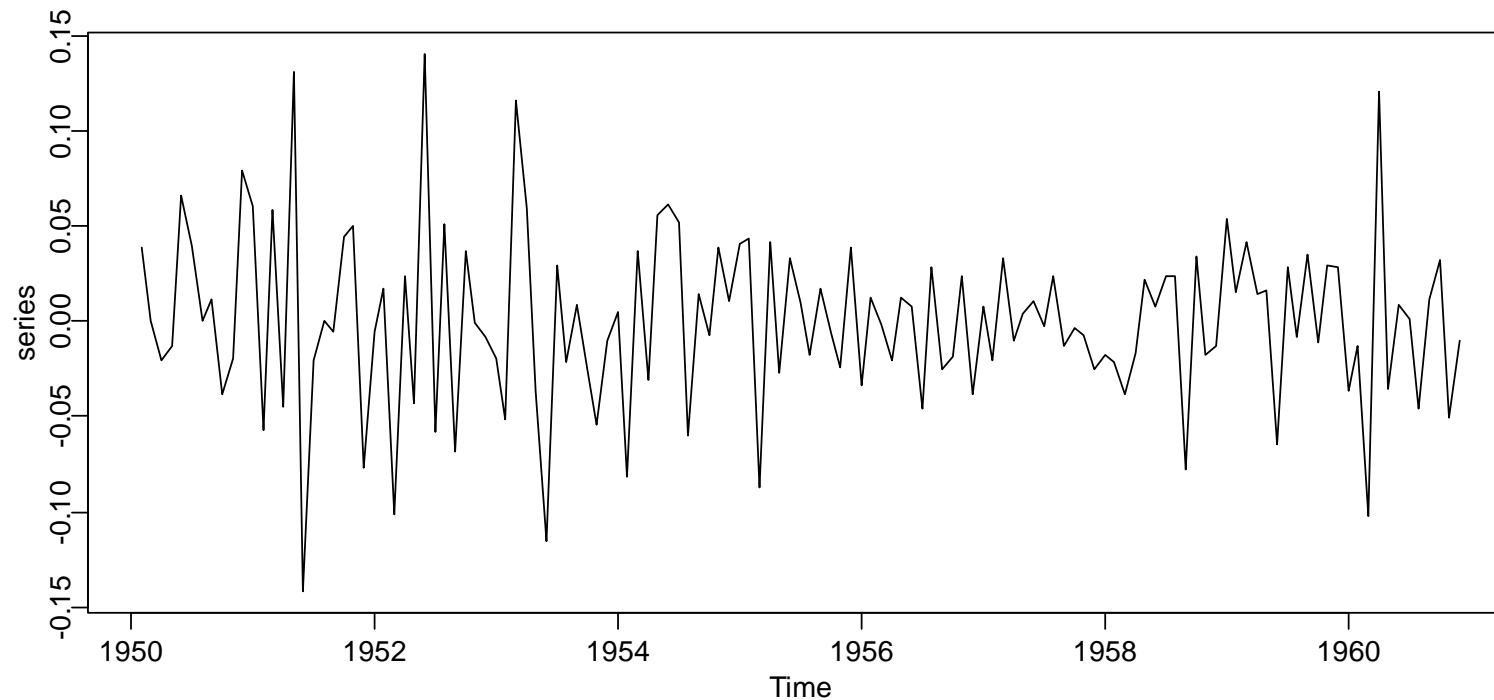
SARIMA(p,d,q)(P,D,Q)^s

We perform some differencing... (→ **see blackboard**)

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ACF/PACF of SARIMA(p,d,q)(P,D,Q)^s



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Modeling the Airline Data

Since there are “big gaps” in ACF/PACF:

$$\begin{aligned} Z_t &= (1 + \beta_1 B)(1 + \gamma_1 B^{12})E_t \\ &= E_t + \beta_1 E_{t-1} + \gamma_1 E_{t-12} + \beta_1 \gamma_1 E_{t-13} \end{aligned}$$

This is an MA(13)-model with many coefficients equal to 0, or equivalently, a SARIMA(0,1,1)(0,1,1)¹².

Note: Every SARIMA(p,d,q)(P,D,Q)^s can be written as an ARMA(p+sP,q+sQ), where many coefficients will be equal to 0.

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SARIMA(p,d,q)(P,D,Q)^s

The general notation is:

$$Z_t = (1 - B)^d (1 - B^s)^D X_t$$

$$\Phi(B)\Phi_s(B^s)Z_t = \Theta(B)\Theta_s(B^s)E_t$$

Interpretation:

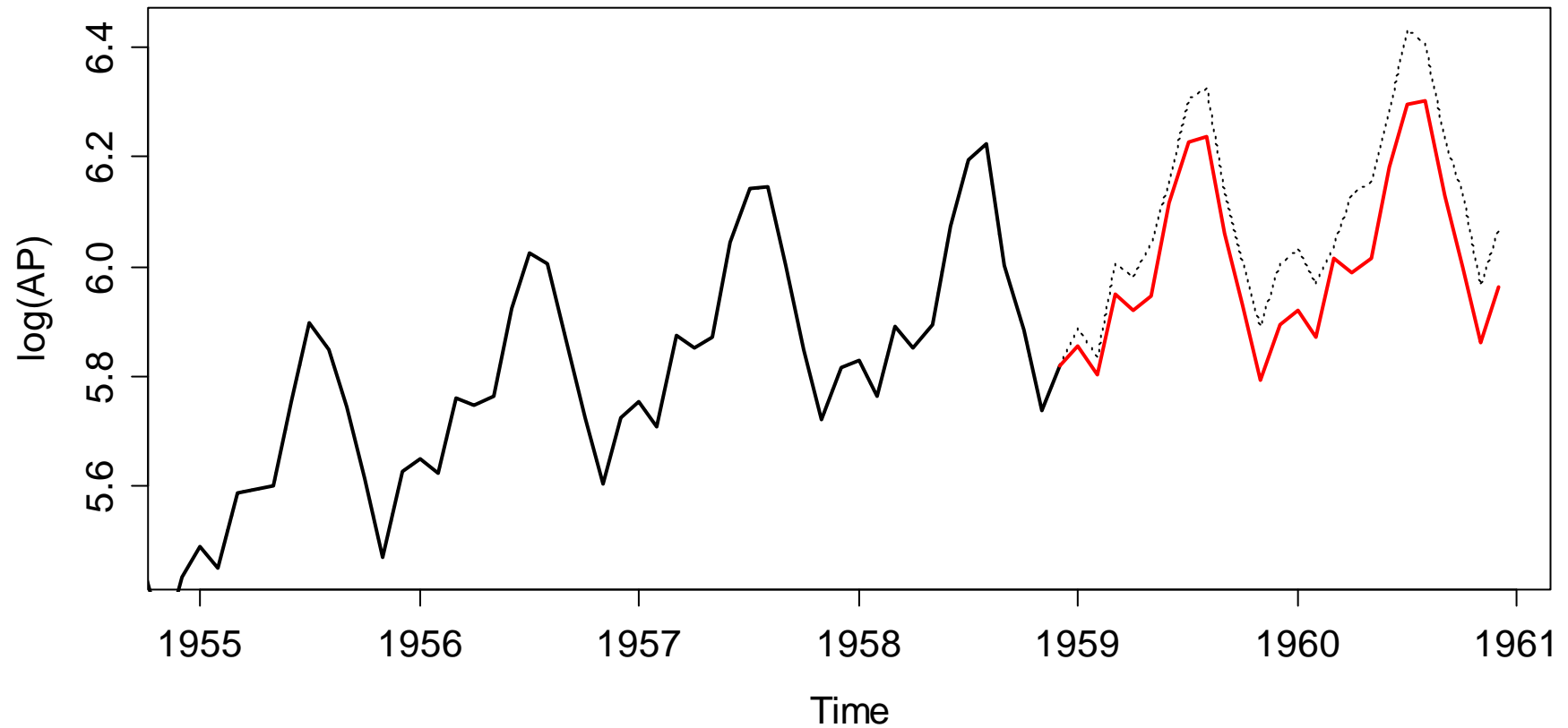
- one typically chooses $d=D=1$
 - s = periodicity in the data (season)
 - P, Q describe the dependency on multiples of the period
- **see blackboard...**

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Forecasting Airline Data

Forecast of $\log(\text{AP})$ with SARIMA(0,1,1)(0,1,1)

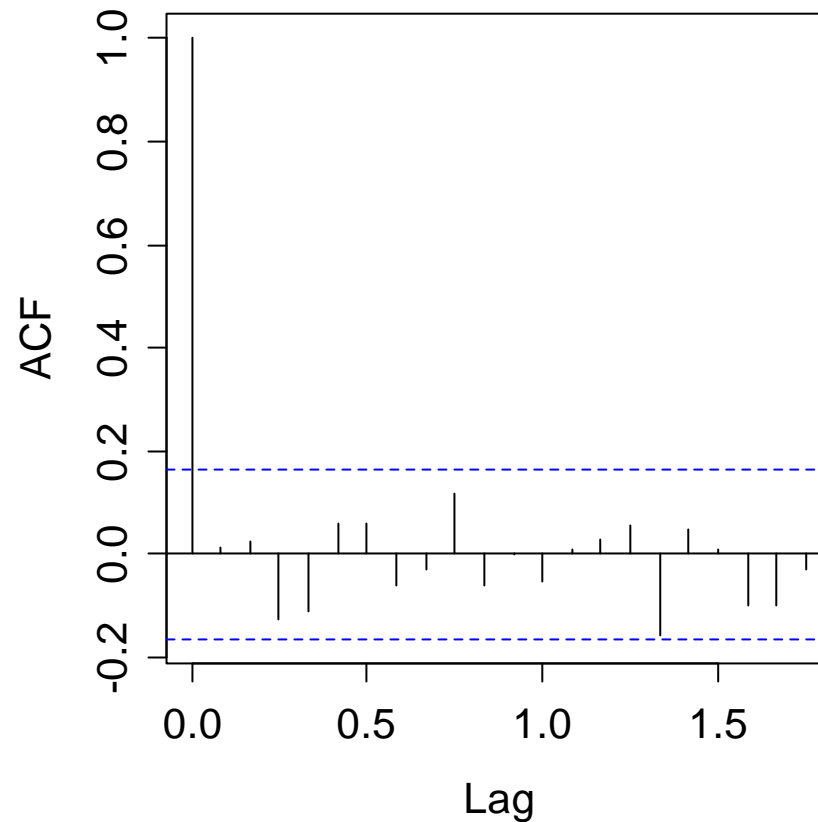


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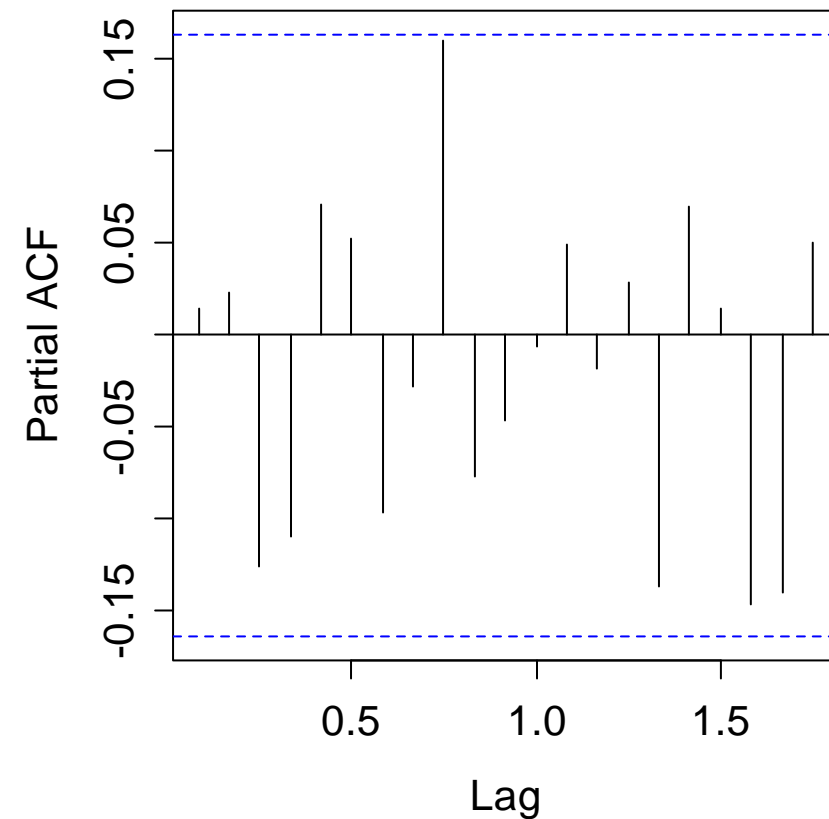
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Residual Analysis of SARIMA(0,1,1)(0,1,1)

ACF



PACF



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Outlook to Non-Linear Models

What are linear models?

Models which can be written as a linear combination of X_t
i.e. all AR-, MA- and ARMA-models

What are non-linear models?

Everything else, e.g. non-linear combinations of X_t ,
terms like X_t^2 in the linear combination, and much more!

Motivation for non-linear models?

- modeling cyclic behavior with quicker increase than decrease
- non-constant variance, even after transforming the series