# Package 'itsmr'

# September 10, 2018

Type Package					
Title Time Series Analysis Using the Innovations Algorithm					
Version 1.9					
<b>Date</b> 2018-09-06					
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<b>Description</b> Provides functions for modeling and forecasting time series data. Forecasting is based on the innovations algorithm. A description of the innovations algorithm can be found in the textbook ``Introduction to Time Series and Forecasting" by Peter J. Brockwell and Richard A. Davis. <a href="http://www.springer.com/us/book/9781475777505">http://www.springer.com/us/book/9781475777505</a> .					
License FreeBSD					
LazyLoad yes					
NeedsCompilation no					
<pre>URL http://eigenmath.org/itsmr-refman.pdf</pre>					
Repository CRAN					
<b>Date/Publication</b> 2018-09-10 18:30:02 UTC					
R topics documented:					
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itsmr-package

Time Series Analysis Using the Innovations Algorithm

# Description

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Provides functions for modeling and forecasting time series data. Forecasting is based on the innovations algorithm. A description of the innovations algorithm can be found in the textbook *Introduction to Time Series and Forecasting* by Peter J. Brockwell and Richard A. Davis.

### **Details**

Package: itsmr
Type: Package
Version: 1.9

Date: 2018-09-06 License: FreeBSD LazyLoad: yes

URL: http://eigenmath.org/itsmr-refman.pdf

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### Author(s)

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#### References

Brockwell, Peter J., and Richard A. Davis. *Introduction to Time Series and Forecasting*. 2nd ed. Springer, 2002.

# **Examples**

```
plotc(wine)
## Define a suitable data model
M = c("log","season",12,"trend",1)
## Obtain residuals and check for stationarity
e = Resid(wine,M)
test(e)
## Define a suitable ARMA model
a = arma(e,p=1,q=1)
## Obtain residuals and check for white noise
ee = Resid(wine,M,a)
test(ee)
## Forecast future values
forecast(wine,M,a)
```

aacvf

Autocovariance of ARMA model

### **Description**

Autocovariance of ARMA model

# Usage

```
aacvf(a, h)
```

### Arguments

a ARMA model
h Maximum lag

4 acvf

### **Details**

The ARMA model is a list with the following components.

phi Vector of AR coefficients (index number equals coefficient subscript)
theta Vector of MA coefficients (index number equals coefficient subscript)
sigma2 White noise variance

# Value

Returns a vector of length h+1 to accomodate lag 0 at index 1.

### See Also

arma

### **Examples**

```
a = arma(Sunspots,2,0)
aacvf(a,40)
```

acvf

Autocovariance of data

### **Description**

Autocovariance of data

### Usage

```
acvf(x, h = 40)
```

# Arguments

x Time series datah Maximum lag

#### Value

Returns a vector of length h+1 to accommodate lag 0 at index 1.

### See Also

plota

```
acvf(Sunspots)
```

ar.inf

airpass

Number of international airline passengers, 1949 to 1960

# Description

Number of international airline passengers, 1949 to 1960

# **Examples**

```
plotc(airpass)
```

ar.inf

Compute AR infinity coefficients

# Description

Compute AR infinity coefficients

# Usage

```
ar.inf(a, n = 50)
```

### **Arguments**

a ARMA model

n Order

### **Details**

The ARMA model is a list with the following components.

phi Vector of AR coefficients (index number equals coefficient subscript)
theta Vector of MA coefficients (index number equals coefficient subscript)
sigma2 White noise variance

### Value

Returns a vector of length n+1 to accommodate coefficient 0 at index 1.

# See Also

ma.inf

6 arar

# **Examples**

```
a = yw(Sunspots,2)
ar.inf(a)
```

arar

Forecast using ARAR algorithm

# Description

Forecast using ARAR algorithm

# Usage

```
arar(y, h = 10, opt = 2)
```

# Arguments

y Time series datah Steps ahead

opt Display option (0 silent, 1 tabulate, 2 plot and tabulate)

# Value

Returns the following list invisibly.

pred Predicted values se Standard errors

1 Lower bounds (95% confidence interval)

u Upper bounds

# See Also

forecast

```
arar(airpass)
```

arma 7

arma

Estimate ARMA model coefficients using maximum likelihood

# Description

Estimate ARMA model coefficients using maximum likelihood

# Usage

```
arma(x, p = 0, q = 0)
```

# Arguments

X	Time series data
р	AR order
q	MA order

### **Details**

Calls the standard R function arima to estimate AR and MA coefficients. The innovations algorithm is used to estimate white noise variance.

### Value

Returns an ARMA model consisting of a list with the following components.

phi	Vector of AR coefficients (index number equals coefficient subscript)
theta	Vector of MA coefficients (index number equals coefficient subscript)
sigma2	White noise variance
aicc	Akaike information criterion corrected
se.phi	Standard errors for the AR coefficients
se.theta	Standard errors for the MA coefficients

### See Also

```
autofit burg hannan ia yw
```

```
M = c("diff",1)
e = Resid(dowj,M)
a = arma(e,1,0)
print(a)
```

8 autofit

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Find the best model from a range of possible ARMA models

# Description

Find the best model from a range of possible ARMA models

# Usage

```
autofit(x, p = 0:5, q = 0:5)
```

### **Arguments**

X	Time series	data	(typically	residuals	from Re	(bize
^	THIE BUILDS	autu	(typicum,	residuals	110111 110	JJ 1 4,

p Range of AR ordersq Range of MA orders

#### **Details**

Tries all combinations of p and q and returns the model with the lowest AICC. The arguments p and q should be small ranges as this function can be slow otherwise. The innovations algorithm is used to estimate white noise variance.

### Value

Returns an ARMA model consisting of a list with the following components.

ph1	Vector of AR coefficients (index number equals coefficient subscript)
theta	Vector of MA coefficients (index number equals coefficient subscript)
sigma2	White noise variance
aicc	Akaike information criterion corrected
se.phi	Standard errors for the AR coefficients
se.theta	Standard errors for the MA coefficients

### See Also

arma

```
M = c("diff",1)
e = Resid(dowj,M)
a = autofit(e)
print(a)
```

burg 9

burg

Estimate AR coefficients using the Burg method

### **Description**

Estimate AR coefficients using the Burg method

### Usage

```
burg(x, p)
```

# Arguments

x Time series data (typically residuals from Resid)

p AR order

#### **Details**

The innovations algorithm is used to estimate white noise variance.

#### Value

Returns an ARMA model consisting of a list with the following components.

phi Vector of AR coefficients (index number equals coefficient subscript)

theta (

sigma2 White noise variance

aicc Akaike information criterion corrected se.phi Standard errors for the AR coefficients

 $\operatorname{se.theta} 0$ 

#### See Also

```
arma hannan ia yw
```

```
M = c("diff",1)
e = Resid(dowj,M)
a = burg(e,1)
print(a)
```

10 deaths

check

Check for causality and invertibility

# Description

Check for causality and invertibility

# Usage

check(a)

# Arguments

а

ARMA model

### **Details**

The ARMA model is a list with the following components.

phi Vector of AR coefficients (index number equals coefficient subscript)
theta Vector of MA coefficients (index number equals coefficient subscript)
sigma2 White noise variance

#### Value

None

# **Examples**

```
a = specify(ar=c(0,0,.99))
check(a)
```

deaths

USA accidental deaths, 1973 to 1978

# Description

USA accidental deaths, 1973 to 1978

```
plotc(deaths)
```

forecast 11

dowj

Dow Jones utilities index, August 28 to December 18, 1972

### **Description**

Dow Jones utilities index, August 28 to December 18, 1972

### **Examples**

```
plotc(dowj)
```

forecast

Forecast future values

# Description

Forecast future values

### Usage

```
forecast(x, M, a, h = 10, opt = 2, alpha = 0.05)
```

### **Arguments**

х	Time series data
М	Data model
а	ARMA model
h	Steps ahead
opt	Display option (0 silent, 1 tabulate, 2 plot and tabulate)
alpha	Level of significance

#### **Details**

The data model can be NULL for none. Otherwise M is a vector of function names and arguments.

Example:

```
M = c("log", "season", 12, "trend", 1)
```

The above model takes the log of the data, then subtracts a seasonal component of period 12, then subtracts a linear trend component.

These are the available functions:

diff Difference the data. Has a single argument, the lag.

hr Subtract harmonic components. Has one or more arguments, each specifying the number of observations per harmo

log Take the log of the data, has no arguments.

season Subtract a seasonal component. Has a single argument, the number of observations per season.

trend Subtract a trend component. Has a single argument, the order of the trend (1 linear, 2 quadratic, etc.)

12 hannan

At the end of the model there is an implicit subtraction of the mean operation. Hence the resulting time series always has zero mean.

All of the functions are inverted before the forecast results are displayed.

### Value

Returns the following list invisibly.

pred	Predicted values
se	Standard errors (not returned for data models with log)
1	Lower bounds (95% confidence interval)
u	Upper bounds

### See Also

```
arma Resid test
```

# **Examples**

```
M = c("log", "season", 12, "trend", 1)
e = Resid(wine, M)
a = arma(e, 1, 1)
forecast(wine, M, a)
```

hannan

Estimate ARMA coefficients using the Hannan-Rissanen algorithm

# Description

Estimate ARMA coefficients using the Hannan-Rissanen algorithm

### Usage

```
hannan(x, p, q)
```

# Arguments

X	Time series data (typically residuals from Resid)
p	AR order
q	MA order $(q > 0)$

### **Details**

The innovations algorithm is used to estimate white noise variance.

hr 13

### Value

Returns an ARMA model consisting of a list with the following components.

phi Vector of AR coefficients (index number equals coefficient subscript)
theta Vector of MA coefficients (index number equals coefficient subscript)
sigma2 White noise variance
aicc Akaike information criterion corrected
se.phi Standard errors for the AR coefficients
se.theta Standard errors for the MA coefficients

#### See Also

```
arma burg ia yw
```

### **Examples**

```
M = c("diff",12)
e = Resid(deaths,M)
a = hannan(e,1,1)
print(a)
```

hr

Estimate harmonic components

### **Description**

Estimate harmonic components

#### Usage

```
hr(x, d)
```

### **Arguments**

x Time series data

d Vector of harmonic periods

### Value

Returns a vector the same length as x. Subtract from x to obtain residuals.

```
y = hr(deaths,c(12,6))
plotc(deaths,y)
```

14 ia

ia

Estimate MA coefficients using the innovations algorithm

### **Description**

Estimate MA coefficients using the innovations algorithm

### Usage

```
ia(x, q, m = 17)
```

### **Arguments**

x Time series data (typically residuals from Resid)

q MA order

m Recursion level

#### **Details**

Normally m should be set to the default value. The innovations algorithm is used to estimate white noise variance.

#### Value

Returns an ARMA model consisting of a list with the following components.

phi 0

theta Vector of MA coefficients (index number equals coefficient subscript)

sigma2 White noise variance

aicc Akaike information criterion corrected

se.phi 0

se. theta Standard errors for the MA coefficients

### See Also

```
arma burg hannan yw
```

```
M = c("diff",1)
e = Resid(dowj,M)
a = ia(e,1)
print(a)
```

ma.inf

lake

Level of Lake Huron, 1875 to 1972

# Description

Level of Lake Huron, 1875 to 1972

# **Examples**

```
plotc(lake)
```

ma.inf

Compute MA infinity coefficients

# Description

Compute MA infinity coefficients

# Usage

```
ma.inf(a, n = 50)
```

### **Arguments**

a ARMA model

n Order

### **Details**

The ARMA model is a list with the following components.

phi Vector of AR coefficients (index number equals coefficient subscript)
theta Vector of MA coefficients (index number equals coefficient subscript)
sigma2 White noise variance

# Value

Returns a vector of length n+1 to accommodate coefficient 0 at index 1.

### See Also

```
ar.inf
```

16 periodogram

### **Examples**

```
M = c("diff",12)
e = Resid(deaths,M)
a = arma(e,1,1)
ma.inf(a,10)
```

periodogram

Plot a periodogram

# Description

Plot a periodogram

# Usage

```
periodogram(x, q = 0, opt = 2)
```

# Arguments

X	Time series data
q	MA filter order
opt	Plot option (0 silent, 1 periodogram only, 2 periodogram and filter)

### **Details**

The filter q can be a vector in which case the overall filter is the composition of MA filters of the designated orders.

### Value

The periodogram vector divided by 2pi is returned invisibly.

### See Also

plots

```
periodogram(Sunspots,c(1,1,1,1))
```

plota 17

plota

Plot data and/or model ACF and PACF

# Description

Plot data and/or model ACF and PACF

# Usage

```
plota(u, v = NULL, h = 40)
```

# **Arguments**

u, v Data and/or ARMA model in either order

h Maximum lag

#### Value

None

### **Examples**

```
plota(Sunspots)
a = yw(Sunspots,2)
plota(Sunspots,a)
```

plotc

Plot one or two time series

# Description

Plot one or two time series

# Usage

```
plotc(y1, y2 = NULL)
```

### Arguments

y1 Data vector (plotted in blue with knots)
y2 Data vector (plotted in red, no knots)

### Value

None

Resid Resid

# **Examples**

```
plotc(uspop)
y = trend(uspop,2)
plotc(uspop,y)
```

plots

Plot spectrum of data or ARMA model

# Description

Plot spectrum of data or ARMA model

# Usage

plots(u)

# Arguments

u

Data vector or an ARMA model

### Value

None

### See Also

```
periodogram
```

# **Examples**

```
a = specify(ar=c(0,0,.99))
plots(a)
```

Resid

Compute residuals

# Description

Compute residuals

# Usage

```
Resid(x, M = NULL, a = NULL)
```

season 19

#### **Arguments**

X	Time series data
М	Data model
а	ARMA model

#### **Details**

The data model can be NULL for none. Otherwise M is a vector of function names and arguments.

Example:

```
M = c("log", "season", 12, "trend", 1)
```

The above model takes the log of the data, then subtracts a seasonal component of period 12, then subtracts a linear trend component.

These are the available functions:

diff Difference the data. Has a single argument, the lag.

hr Subtract harmonic components. Has one or more arguments, each specifying the number of observations per harmo log Take the log of the data, has no arguments.

season Subtract a seasonal component. Has a single argument, the number of observations per season.

trend Subtract a trend component. Has a single argument, the order of the trend (1 linear, 2 quadratic, etc.)

At the end of the model there is an implicit subtraction of the mean operation. Hence the resulting time series always has zero mean.

#### Value

Returns a vector of residuals the same length as x.

#### See Also

test

# **Examples**

```
M = c("log", "season", 12, "trend", 1)
e = Resid(wine, M)

a = arma(e, 1, 1)
ee = Resid(wine, M, a)
```

season

Estimate seasonal component

### **Description**

Estimate seasonal component

20 selftest

### Usage

```
season(x, d)
```

# Arguments

x Time series data

d Number of observations per season

### Value

Returns a vector the same length as x. Subtract from x to obtain residuals.

#### See Also

trend

# **Examples**

```
y = season(deaths,12)
plotc(deaths,y)
```

selftest

Run a self test

# Description

Run a self test

# Usage

```
selftest()
```

### **Details**

This function is a useful check if the code is modified.

# Value

None

```
selftest()
```

smooth.exp 21

sim

Generate synthetic observations

# Description

Generate synthetic observations

# Usage

```
sim(a, n = 100)
```

### **Arguments**

a ARMA model

n Number of synthetic observations required

### **Details**

The ARMA model is a list with the following components.

phi Vector of AR coefficients (index number equals coefficient subscript)
theta Vector of MA coefficients (index number equals coefficient subscript)
sigma2 White noise variance

#### Value

Returns a vector of n synthetic observations.

### **Examples**

```
a = specify(ar=c(0,0,.99))
x = sim(a,60)
plotc(x)
```

smooth.exp

Apply an exponential filter

# Description

Apply an exponential filter

### Usage

```
smooth.exp(x, alpha)
```

22 smooth.fft

### **Arguments**

x Time series data

alpha Smoothness setting, 0-1

#### **Details**

Zero is maximum smoothness.

#### Value

Returns a vector of smoothed data the same length as x.

# **Examples**

```
y = smooth.exp(strikes,.4)
plotc(strikes,y)
```

smooth.fft

Apply a low pass filter

### **Description**

Apply a low pass filter

### Usage

```
smooth.fft(x, f)
```

# **Arguments**

x Time series data

f Cut-off frequency, 0-1

### **Details**

The cut-off frequency is specified as a fraction. For example, c=.25 passes the lowest 25% of the spectrum.

#### Value

Returns a vector the same length as x.

```
y = smooth.fft(deaths,.1)
plotc(deaths,y)
```

smooth.ma 23

smooth.ma

Apply a moving average filter

# Description

Apply a moving average filter

# Usage

```
smooth.ma(x, q)
```

# Arguments

x Time series data q Filter order

# **Details**

The averaging function uses 2q+1 values.

#### Value

Returns a vector the same length as x.

# **Examples**

```
y = smooth.ma(strikes,2)
plotc(strikes,y)
```

 ${\tt smooth.rank}$ 

Apply a spectral filter

# Description

Apply a spectral filter

# Usage

```
smooth.rank(x, k)
```

### **Arguments**

x Time series data

k Number of frequencies

24 specify

### **Details**

Passes the mean and the k frequencies with the highest amplitude. The remainder of the spectrum is filtered out.

### Value

Returns a vector the same length as x.

# **Examples**

```
y = smooth.rank(deaths,2)
plotc(deaths,y)
```

specify

Specify an ARMA model

# Description

Specify an ARMA model

### Usage

```
specify(ar = 0, ma = 0, sigma2 = 1)
```

# Arguments

ar	Vector of AR coefficients (index number equals coefficient subscript)
ma	Vector of MA coefficients (index number equals coefficient subscript)
sigma2	White noise variance

# Value

Returns an ARMA model consisting of a list with the following components.

phi	Vector of AR coefficients (index number equals coefficient subscript)
theta	Vector of MA coefficients (index number equals coefficient subscript)
sigma2	White noise variance

```
specify(ar=c(0,0,.99))
```

strikes 25

strikes

USA union strikes, 1951-1980

# Description

USA union strikes, 1951-1980

### **Examples**

```
plotc(strikes)
```

Sunspots

Number of sunspots, 1770 to 1869

# Description

Number of sunspots, 1770 to 1869

# **Examples**

plotc(Sunspots)

test

Test residuals for stationarity and randomness

### **Description**

Test residuals for stationarity and randomness

# Usage

test(e)

# Arguments

е

Time series data (typically residuals from Resid)

### **Details**

Plots ACF, PACF, residuals, and QQ. Displays results for Ljung-Box, McLeod-Li, turning point, difference-sign, and rank tests. The plots can be used to check for stationarity and the other tests check for white noise.

26 trend

### Value

None

#### See Also

Resid

# **Examples**

```
M = c("log", "season", 12, "trend", 1)
e = Resid(wine, M)
test(e) ## Is e stationary?
a = arma(e, 1, 1)
ee = Resid(wine, M, a)
test(ee) ## Is ee white noise?
```

trend

Estimate trend component

# Description

Estimate trend component

### Usage

```
trend(x, p)
```

# Arguments

x Time series data

p Polynomial order (1 linear, 2 quadratic, etc.)

#### Value

Returns a vector the same length as x. Subtract from x to obtain residuals. The returned vector is the least squares fit of a polynomial to the data.

### See Also

season

```
y = trend(uspop,2)
plotc(uspop,y)
```

wine 27

wine

Australian red wine sales, January 1980 to October 1991

# Description

Australian red wine sales, January 1980 to October 1991

### **Examples**

```
plotc(wine)
```

уw

Estimate AR coefficients using the Yule-Walker method

# Description

Estimate AR coefficients using the Yule-Walker method

### Usage

```
yw(x, p)
```

#### **Arguments**

x Time series data (typically residuals from Resid)

p AR order

#### **Details**

The innovations algorithm is used to estimate white noise variance.

#### Value

Returns an ARMA model consisting of a list with the following components.

phi Vector of AR coefficients (index number equals coefficient subscript)

theta 0

sigma2 White noise variance

aicc Akaike information criterion corrected se.phi Standard errors for the AR coefficients

 ${\tt se.theta} \qquad \quad 0$ 

#### See Also

arma burg hannan ia

28 yw

```
M = c("diff",1)
e = Resid(dowj,M)
a = yw(e,1)
```

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