

Package ‘ftnonpar’

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Title Features and Strings for Nonparametric Regression

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Description The package contains R-functions to perform the methods in nonparametric regression and density estimation, described in Davies, P. L. and Kovac, A. (2001) Local Extremes, Runs, Strings and Multiresolution (with discussion) Annals of Statistics. 29. p1-65 Davies, P. L. and Kovac, A. (2004) Densities, Spectral Densities and Modality Annals of Statistics. Annals of Statistics. 32. p1093-1136 Kovac, A. (2006) Smooth functions and local extreme values. Computational Statistics and Data Analysis (to appear) D\{"umbgen, L. and Kovac, A. (2006) Extensions of smoothing via taut strings Davies, P. L. (1995) Data features. Statistica Neerlandica 49,185-245.

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balloon	<i>Data from a weather balloon</i>
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Description

The data consist of 4984 observations taken from a balloon about 30 kilometres above the surface of the earth. The outliers are caused by the fact that the balloon slowly rotates, causing the ropes from which the measuring instrument is suspended to cut off the direct radiation from the sun.

Usage

```
data(djdata)
```

Format

A one-dimensional data set of size 4984

Source

Davies, L. and Gather, U. (1993), The Identification of Multiple Outliers. *JASA* **88**, 782–801.

Examples

```
data(balloon)
plot(balloon)
```

dclaw *The Claw Distribution*

Description

Generates a sample from the claw distribution.

Usage

```
dclaw(x)
```

Arguments

x Vector of points where the claw density is evaluated.

Author(s)

Arne Kovac <A.Kovac@bristol.ac.uk>

References

Marron, J. S. and Wand, M. P. (1992) Exact mean integrated squared error

See Also

[rclaw](#)

Examples

```
plot(dclaw(seq(-3,3,len=1000)), ty="l")
```

djdata *Donoho-Johnstone test signals*

Description

Four samples, each of size 2048 from Donoho and Johnstone's famous signals commonly used as test beds for smoothing methods.

Usage

```
data(djdata)
```

Format

Four one-dimensional data sets, each of size 2048

Source

Donoho, D. L. et al (1995) Wavelet Shrinkage: Asymptopia? *Journal of the Royal Statistical Society series B*, **57**, 301–337.

Examples

```
data(djdata)
par(mfrow=c(2,2))
plot(djdoppler)
plot(djheavisine)
plot(djblocks)
plot(djbumps)
```

frun

Runs and local extremes

Description

Calculation of bounds for functions such that the residuals satisfy a run criterion

Usage

```
frun(y, ..., alpha = 0.5, r = 0, mr = 0)
```

Arguments

y	the data
...	an optional argument which specifies the approximate positions of the local extreme values. These should be consistent with the run length otherwise the result will be nonsensical. Should you wish to use this option then you should first run the macro without it. The item xb of the output list gives the acceptable limits of the local extreme values. You can then specify the positions within these limits.
alpha	Quantile determining the acceptable run length.
r	Acceptable run length: Overrides alpha if not 0
mr	mr=0 minimizes the run length consistent with the number of local extreme values found for the specified run length. mr=1 disables the option.

Value

l1	lower bounds
u1	upper bounds
l2	lower bound with specified extremes: the default choices for the positions of the local extreme values are the mid-points of the intervals specified by xb above.
u2	upper bound with specified extremes
f	function between l2 and u2 satisfying run condition

xb	bounds for location of extremes: the position of the i th extreme value lies between $xb[2*i-1]$ and $xb[2*i]$
nx	number of extremes
r	run length: may differ from specified if $mr=1$

Note

IN GENERAL THE MEAN OF THE BOUNDS l_2 AND u_2 $(l_2+u_2)/2$ GIVES A BETTER REGRESSION FUNCTION THAN f . HOWEVER THIS FUNCTION IS INFINITE AT THE TWO ENDPOINTS AND AT LOCAL EXTREME VALUES. IN THESE INTERVALS IT CAN BE REPLACED BY ANY VALUES WHICH DO NOT ALTER THE NUMBER OF LOCAL EXTREME VALUES. THE MEDIAN OF THE y -VALUES IN THESE INTERVALS IS A REASONABLE CHOICE.

Author(s)

Laurie Davies <Laurie.Davies@uni-essen.de>

References

Davies, P. L. (1995) Data features. *Statistica Neerlandica* 49,185-245.

Davies, P. L. and Kovac, A. (2001) Local Extremes, Runs, Strings and Multiresolution (with discussion) *Annals of Statistics*. 29. p1-65

See Also

[mintvmon](#), [pmreg](#), [l1pmreg](#)

genpmreg

Piecewise monotone regression with generalised taut strings

Description

Applies the generalised taut string method to one-dimensional data.

Usage

```
genpmreg(y, beta = 0.5, squeezing.factor = 0.5, verbose = FALSE, localsqueezing = TRUE, DYADIC = TRUE,
```

Arguments

y	observed values (ordered by value of independent variable)
beta	If method=1 specifies the quantile
squeezing.factor	The amount of decrement applied to the bandwidths
verbose	logical, if T progress (for each iteration) is illustrated graphically

localsqueezing	logical, if T (default) the bandwidth is changed locally.
DYADIC	If T checks the multiresolution criterion only on dyadic intervals, otherwise all intervals are checked
thr.const	smoothing parameter for the multiresolution criterion (should be approximately 2.5)
extrema.nr	if set to a positive integer an approximation with the specified number of local extreme values is calculated
bandwidth	if set to a positive value the specified bandwidth is used instead of the multiresolution criterion.
SETTOMEAN	logical, if T (default) the value of the taut string approximation at local extreme values is set to the mean or median of the observations on the interval where the extremum is taken.
method	The method used which can be 1 (quantile regression), 2 (usual taut string), 3 (logistic regression) and 4 (Poisson regression)
...	Passed to the plot command if verbose=T

Value

A list with components

y	The approximation of the given data
lambda	Values for lambda used
nmax	Number of local extreme values

Author(s)

Arne Kovac <A.Kovac@bristol.ac.uk>

References

DUmbgen, L. and Kovac, A. (2006) Extensions of smoothing via taut strings

See Also

[pmreg](#)

Examples

```
data(djdata)
par(mfrow=c(2,2))
plot(djblocks,col="grey")
lines(genpmreg(djblocks,verbose=FALSE,method=2,thr.const=2.5)$y,col="red")
title("Usual taut string method")
ind <- sample(1:length(djblocks),300)
djblocks[ind] <- djblocks[ind]+rnorm(length(ind),0,100)
plot(djblocks,col="grey")
lines(genpmreg(djblocks,verbose=FALSE,method=2)$y,col="red")
title("Usual taut string method with outliers")
```

```
plot(djblocks,col="grey")
lines(genpmreg(djblocks,verbose=FALSE,method=1)$y,col="red")
title("Robust taut string method with outliers")
plot(djblocks,col="grey",ylim=c(-10,10))
lines(genpmreg(djblocks,verbose=FALSE,method=1)$y,col="red")
title("Again robust taut string method, different scaling of y-axis")
```

kkuip

Generalized Kuiper Metrics

Description

Calculates the generalized k-th Kuiper metric

Usage

```
kkuip(x, k=1)
```

Arguments

x	Data vector
k	Number of intervals to be used for the generalized Kuiper metric. The usual Kuiper metric is obtained by the default value 1.

Value

metric	Value of the metric
a	The left borders of the intervals where the maximum is obtained.
b	The right borders of the intervals where the maximum is obtained.

Author(s)

Arne Kovac <A.Kovac@bristol.ac.uk>

References

Davies, P. L. and Kovac, A. (2001) Densities, Spectral Densities and Modality

See Also

[pmden](#)

Examples

```
aaa <- rclaw(500)
kkuip(aaa,9)
```

l1pmreg

*Piecewise monotone nonparametric quantile regression***Description**

Applies the generalized taut string method to quantile regression.

Usage

```
quantpmreg(y, beta = 0.5, squeezing.factor = 0.5, verbose = FALSE, localsqueezing = TRUE, DYADIC = TRUE)
l1pmreg(y, beta=0.5, squeezing.factor = 0.5, verbose = FALSE, localsqueezing = TRUE, DYADIC = TRUE, th
```

Arguments

y	observed values (ordered by value of independent variable)
beta	quantile. The default is 0.5 which corresponds to the robust taut string.
squeezing.factor	The amount of decrement applied to the bandwidths
verbose	logical, if T progress (for each iteration) is illustrated graphically
localsqueezing	logical, if T (default) the bandwidth is changed locally.
DYADIC	logical, if T (default) the multiresolution criterion is only verified on intervals with dyadic endpoints.
thr.const	smoothing parameter for the multiresolution criterion (should be approximately 2)
extrema.nr	if set to a positive integer an approximation with the specified number of local extreme values is calculated
bandwidth	if set to a positive value the specified bandwidth is used instead of the multiresolution criterion.
SETTOMEAN	logical, if T (default) the value of the taut string approximation at local extreme values is set to the mean or median of the observations on the interval where the extremum is taken.
method	The method used which can be 1 (quantile regression), 2 (usual taut string), 3 (logistic regression) and 4 (Poisson regression)
...	Passed to the plot command if verbose=T

Value

A list with components

y	The approximation of the given data
lambda	The final values of lambda
nmax	Number of local extreme values

Author(s)

Arne Kovac <A.Kovac@bristol.ac.uk>

References

D'Umbgen, L. and Kovac, A. (2003) Extensions of smoothing via taut strings

See Also

[pmreg](#), [frun](#)

mintvmon

Minimization of total variation

Description

Finds a function vector which minimizes the total variation of the function or a derivative under multiresolution constraints and monotonicity and convexity constraints.

Usage

```
mintvmon(y, sigma = -1, DYADIC = TRUE, thresh = -1, method = 2, MONCONST = TRUE, CONVCONST = FALSE)
```

Arguments

y	observed values (ordered by value of independent variable).
sigma	if set to a positive value the standard deviation is set to sigma and not estimated from the data
DYADIC	logical, if T (default) the multiresolution constraints are only verified on intervals with dyadic endpoints
thresh	if set to a positive value other thresholds for the multiresolution criterion than the default $\sqrt{2 \cdot \log(n)} \cdot \text{sigma}$ can be used.
method	Number of derivative the total variation of which is minimized. Possible values are 0,1,2. Higher values lead to numerical inconsistencies.
MONCONST	logical, if T (default) additional monotonicity constraints are gathered from minimizing the total variation of f. Makes only sense, if method is 1 or 2.
CONVCONST	logical, if T (default) additional convexity constraints are gathered from minimizing the total variation of f'. Makes only sense, if method is 2.

Value

A list with components

<code>y</code>	The approximation of the given data
<code>derivsign</code>	Vector of 1 and -1, monotonicity constraints used if <code>MONCONST</code> was true
<code>secsign</code>	Vector of 1 and -1, convexity constraints used if <code>CONVCONST</code> was true
<code>jact</code>	Left endpoints of active multiresolution constraints for the final approximation
<code>kact</code>	Right endpoints of active multiresolution constraints for the final approximation
<code>signact</code>	Vector of 1 and -1, gives for each active multiresolution constraints, if the residuals on that interval attain upper or lower bound
<code>pl</code>	Left endpoint of piecewise constant intervals of the derivative of f being minimized
<code>pr</code>	Right endpoint of piecewise constant intervals of the derivative of f being minimized

Author(s)

Arne Kovac

References

Kovac, A. (2003) Minimizing Total Variation under Multiresolution Conditions

See Also

[pmreg](#)

Examples

```
data(djdata)
djdoppler.tv0 <- mintvmon(djdoppler,method=0)
djdoppler.tv1 <- mintvmon(djdoppler,method=1)
djdoppler.tv2 <- mintvmon(djdoppler)
par(mfrow=c(2,2))
plot(djdoppler,col="lightgrey")
plot(djdoppler,col="lightgrey")
lines(djdoppler.tv0$y,col="blue")
plot(djdoppler,col="lightgrey")
lines(djdoppler.tv1$y,col="green")
plot(djdoppler,col="lightgrey")
lines(djdoppler.tv2$y,col="red")
```

pmden *Piecewise monotone density estimation with taut strings*

Description

Applies the taut string method to one-dimensional data.

Usage

```
pmden(x, DISCR=FALSE, verbose = FALSE, bandwidth=-1, extrema.nr = -1, accuracy = mad(x)/1000, extrema.
```

Arguments

x	observed values
DISCR	logical, if T a discrete density is fitted
verbose	logical, if T progress (for each iteration) is illustrated graphically
bandwidth	if set to a positive value the specified bandwidth is used instead of the automatic criterion based on generalized Kuiper metrics.
extrema.nr	if set to a positive integer an approximation with the specified number of local extreme values is calculated
accuracy	Precision of the data. Handling of identical observations depends on this parameter.
extrema.mean	logical, if T the value at the local extrema is changed to the mean frequency of observations on that interval
maxkuipnr	The order of the highest generalized Kuiper metric used for the automatic choice of the bandwidth
asymptbounds	If set to T asymptotic bounds derived from a Brownian Bridge are used for the Kuiper criterion. Otherwise simulated bounds for various sample sizes are interpolated for the size of the data x
tolerance	Accuracy used for the determination of the bandwidth when extrema.nr is greater than 0.
localsq	If set to TRUE (default) performs, if necessary, additional local squeezing after the Kuiper metrics are small enough
locsq.factor	The amount of decrement applied to the bandwidths if local squeezing is carried out.

Value

y	values of the density approximation between the observations
widthes	bandwidth used for the taut string approximation
nmax	number of local extreme values
ind	indices of knots points of the taut string
trans	taut string at the observations, should look like uniform noise

Author(s)

Arne Kovac <A.Kovac@bristol.ac.uk>

References

Davies, P. L. and Kovac, A. (2003) Densities, Spectral Densities and Modality

See Also

[pmreg](#), [l1pmreg](#), [pmspec](#)

Examples

```
aaa <- rclaw(500)
pmden(aaa, verb=TRUE)$n
```

pmlogreg

Piecewise monotone logistic regression with taut strings

Description

Applies the taut string method to binary data.

Usage

```
pmlogreg(y, thr.const=2.5, verbose=FALSE, extrema.nr=-1, bandwidth=-1,
localsqueezing=TRUE, squeezing.factor=0.5, tolerance=0.001, extrema.mean=TRUE)
```

Arguments

y	observed values (ordered by value of independent variable)
thr.const	smoothing parameter for the multiresolution criterion (should be approximately 2.5)
verbose	logical, if T progress (for each iteration) is illustrated graphically
extrema.nr	if set to a positive integer an approximation with the specified number of local extreme values is calculated
bandwidth	if set to a positive value the specified bandwidth is used instead of the multiresolution criterion.
localsqueezing	logical, if T (default) the bandwidth is changed locally.
squeezing.factor	The amount of decrement applied to the bandwidths
tolerance	Accuracy used for the determination of the bandwidth when extrema.nr is greater than 0.
extrema.mean	logical, if T (default) the value of the taut string approximation at local extreme values is set to the mean of the observations on the interval where the extremum is taken.

Value

A list with components

y	The approximation of the given data
widthes	Bandwidth used
nmax	Number of local extreme values
knotsind	Indices of knot points
knotsy	y-koordinates of knots of the taut string

Author(s)

Arne Kovac <A.Kovac@bristol.ac.uk>

See Also

[l1pmreg](#), [pmden](#), [pmspec](#)

Examples

```
aaa<-rbinom(1024,1,0.5+0.5*sin(seq(0,10*pi,len=1024)))
pmlogreg(aaa,verbose=TRUE)$n
```

pmreg

Piecewise monotone regression with taut strings

Description

Applies the taut string method to one-dimensional data.

Usage

```
pmreg(y, thr.const=2.3, verbose=FALSE, extrema.nr=-1, bandwidth=-1,
sigma=-1, localsqueezing=TRUE, squeezing.factor=0.5, tolerance=1e-08,
extrema.mean=TRUE,DYADIC=TRUE,dyad.factor=1.1,POSTISO=TRUE)
```

Arguments

y	observed values (ordered by value of independent variable)
thr.const	smoothing parameter for the multiresolution criterion (should be approximately 2.3)
verbose	logical, if T progress (for each iteration) is illustrated graphically
extrema.nr	if set to a positive integer an approximation with the specified number of local extreme values is calculated
bandwidth	if set to a positive value the specified bandwidth is used instead of the multiresolution criterion.

sigma	if set to a positive value sigma the standard deviation is set to sigma and not estimated from the data
localsqueezing	logical, if TRUE (default) the bandwidth is changed locally.
squeezing.factor	The amount of decrement applied to the bandwidths
tolerance	Accuracy used for the determination of the bandwidth when extrema.nr is greater than 0.
extrema.mean	logical, if TRUE (default) the value of the taut string approximation at local extreme values is set to the mean of the observations on the interval where the extremum is taken.
DYADIC	If TRUE the multiresolution constraints are only checked on dyadic intervals.
dyad.factor	If the multiresolution constraints are checked on dyadic intervals, dyad.factor determines the ratio between the lengths of two subsequent level (default is 1.1).
POSTISO	If TRUE (default) any bias caused by local squeezing is removed by applying isotonic and isotonic regression between each two local extreme values.

Value

A list with components

y	The approximation of the given data
sigma	Standard deviation used
widthes	Bandwidth used
nmax	Number of local extreme values
knotsind	Indices of knot points
knotsy	y-kordinates of knots of the taut string

Author(s)

Arne Kovac <A.Kovac@bristol.ac.uk>

References

Davies, P. L. and Kovac, A. (2001) Local Extremes, Runs, Strings and Multiresolution (with discussion) *Annals of Statistics*. 29. p1-65

See Also

[mintvmon](#), [l1pmreg](#), [pmden](#), [pmspec](#)

Examples

```
data(djdata)
pmreg(djdoppler, verbose=TRUE)$n
```

pmspec

*Piecewise monotone spectral density approximation with taut strings***Description**

Applies the taut string method to spectral densities.

Usage

```
pmspec(x, pks=0, alpha=0.9, sqzf=0.9, mult=0, lcl=0, ln=0, fig = 0, pow=10^-2)
```

Arguments

x	data
pks	if pks=0 then the number of peaks is determined automatically. If pks = k > 0 then a density with n paeks is returned pks
alpha	If pks=0 then on Gaussian test beds the number of peaks of the true density is at least the returned value with an asymptotic probability of at least alpha. If pks > 0 then the parameter is not operational alpha
sqzf	Squeeze factor for the taut string sqzf
mult	If mult = T then rescaled empirical density is compared with the exponential distribution only on intervals forming a multiresolution scheme. If T = F then all intervals are used. If the sample size of the data is less than 512 then T = F is default mult
lcl	If lcl = T then string is calculated using local squeezing. If lcl = F then global squeezing is used.lcl
fig	If fig = T then the data are automticall plotted fig
ln	If fig = T and ln = T then the densities are plotted on a log scale. If ln = F then no transformation is performed ln
pow	Peaks with a power less than pow * total power will be ignored

Value

edf	Emprical spectral density
df	String density
pks	Number of peaks
ll	Lower boundary for string
uu	Upper boundary for string
str	String

Author(s)

Laurie Davies <Laurie.Davies@uni-essen.de>

References

Davies, P.L. and Kovac, A. (2001) Local extremes, runs, strings and multiresolution, *Annals of Statistics* 29 (1) 1-65. Davies, P. L. and Kovac, A. (2002) Densities, Spectral Densities and Modality

See Also

[pmreg](#), [l1pmreg](#), [pmden](#)

rclaw

The Claw Distribution

Description

Generates a sample from the claw distribution.

Usage

```
rclaw(n)
```

Arguments

n Sample size

Author(s)

Arne Kovac <A.Kovac@bristol.ac.uk>

References

Marron, J. S. and Wand, M. P. (1992) Exact mean integrated squared error

See Also

[dclaw](#)

Examples

```
aaa <- rclaw(500)
kkuip(aaa, 9)
```

rtennormal	<i>Mixture of ten normal distributions</i>
------------	--

Description

Generates a sample from a mixture of ten normal distributions.

Usage

```
rtennormal(n)
```

Arguments

n	Sample size
---	-------------

Author(s)

Arne Kovac <A.Kovac@bristol.ac.uk>

See Also

[rclaw](#)

Examples

```
aaa <- rtennormal(500)
kkuip(aaa,9)
```

smdenreg	<i>Piecewise monotone density estimation with smooth taut strings</i>
----------	---

Description

Applies the smooth taut string method to one-dimensional data.

Usage

```
smdenreg(x, verbose = FALSE, bandwidth=-1, maxkuiplr=19, asympbounds=FALSE,
squeezing.factor=0.9, firstlambda=10, smqeps=1/length(x), fsign=double(0),
gensign=TRUE,...)
```

Arguments

x	observed values
verbose	logical, if T progress (for each iteration) is illustrated graphically
bandwidth	if set to a positive value the specified bandwidth is used instead of the automatic criterion based on generalized Kuiper metrics.
maxkuipnr	The order of the highest generalized Kuiper metric used for the automatic choice of the bandwidth
asymptbounds	If set to T asymptotic bounds derived from a Brownian Bridge are used for the Kuiper criterion. Otherwise simulated bounds for various sample sizes are interpolated for the size of the data x
squeezing.factor	The amount of decrement applied to the bandwidths
firstlambda	Initial value of lambda's global squeezing.
smqeps	Distance between the (equally-spaced) time points.
fsign	Monotonicity constraints, vector of size n-1 of -1,0 and 1's. If fsign[i]==1, then fhat[i+1]>= fhat[i]. If fsign[i]==-1, then fhat[i+1]<=f[i]. Otherwise no constraint at this position.
gensign	If TRUE the taut string method is used to automatically produce suitable monotonicity constraints.
...	Passed to the plot command if verbose=T.

Value

x	The sorted data
y	values of the density approximation between the observations
nmax	Number of local extreme values
trans	taut string at the observations, should look like uniform noise

Author(s)

Arne Kovac <A.Kovac@bristol.ac.uk>

References

Kovac, A. (2006) Smooth functions and local extreme values. Technical Report

See Also

[pmreg](#), [l1pmreg](#), [pmspec](#)

Examples

```
y <- rclaw(500)
hist(y,col="lightgrey",40,freq=FALSE)
lines(smdenreg(y),col="red")
```

smqreg *Smooth piecewise monotone regression with taut strings*

Description

Applies the smooth taut string method to one-dimensional data.

Usage

```
smqreg(y, thr.const=2.5, verbose=FALSE, bandwidth=-1,
sigma=-1, localsqueezing=TRUE, squeezing.factor=0.5, DYADIC=TRUE,
firstlambda=100, smqeps=1/length(y), fsign=double(0), gensign=TRUE,
tolerance = 1e-12, ...)
```

Arguments

y	observed values (ordered by value of independent variable)
thr.const	smoothing parameter for the multiresolution criterion (should be approximately 2.5)
verbose	logical, if T progress (for each iteration) is illustrated graphically
bandwidth	if set to a positive value the specified bandwidth is used instead of the multiresolution criterion.
sigma	if set to a positive value sigma the standard deviation is set to sigma and not estimated from the data
localsqueezing	logical, if T (default) the bandwidth is changed locally.
squeezing.factor	The amount of decrement applied to the bandwidths
DYADIC	If TRUE the multiresolution constraints are only checked on dyadic intervals.
firstlambda	Initial value of lambda's for local or global squeezing.
smqeps	Distance between the (equally-spaced) time points.
fsign	Monotonicity constraints, vector of size n-1 of -1,0 and 1's. If fsign[i]==1, then fhat[i+1]>= fhat[i]. If fsign[i]==-1, then fhat[i+1]<=f[i]. Otherwise no constraint at this position.
gensign	If TRUE the taut string method is used to automatically produce suitable monotonicity constraints.
tolerance	Precision for the nested intervals for solving the minimisation problem.
...	Passed to the plot command if verbose=T.

Value

A list with components

y	The approximation of the given data
nmax	Number of local extreme values
sigma	Standard deviation used

Author(s)

Arne Kovac <A.Kovac@bristol.ac.uk>

References

Kovac, A. (2006) Smooth functions and local extreme values. Technical Report

See Also

[pmreg](#), [mintvmon](#), [l1pmreg](#), [pmden](#), [pmspec](#)

Examples

```
data(djdata)
par(mfrow=c(2,2))
plot(djblocks,col="grey")
lines(smqreg(djblocks)$y,col="red")
plot(djbumps,col="grey")
lines(smqreg(djbumps)$y,col="red")
plot(djheavisine,col="grey")
lines(smqreg(djheavisine)$y,col="red")
plot(djdoppler,col="grey")
lines(smqreg(djdoppler)$y,col="red")
```

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