## Package: wavelets (via r-universe)

September 7, 2024

Version 0.3-0.2

Date 2020-02-16

Title Functions for Computing Wavelet Filters, Wavelet Transforms and Multiresolution Analyses

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**Depends**  $R$  ( $>= 2.10$ ), methods

Description Contains functions for computing and plotting discrete wavelet transforms (DWT) and maximal overlap discrete wavelet transforms (MODWT), as well as their inverses. Additionally, it contains functionality for computing and plotting wavelet transform filters that are used in the above decompositions as well as multiresolution analyses.

LazyLoad yes

License GPL  $(>= 2)$ 

NeedsCompilation yes

Date/Publication 2020-02-17 19:40:02 UTC

Repository https://ealdrich.r-universe.dev

RemoteUrl https://github.com/cran/wavelets

RemoteRef HEAD

RemoteSha 8365ac9d4a96b9286133ad5f255418163c1746c3

## **Contents**



#### <span id="page-1-0"></span> $2$  align



<span id="page-1-1"></span>align *Align Wavelet Transform Coefficients*

## Description

Aligns both wavelet (high pass) and scaling (low pass) coefficients of objects of class dwt, modwt, dwpt, and modwpt using phase shift values computed by wt.filter.shift.

## Usage

align(wt, coe=FALSE, inverse=FALSE)

## Arguments



#### <span id="page-2-0"></span> $dwt$  3

## Details

An object of class dwt, modwt, is characterized as 'unaligned' if the value in the aligned slot is FALSE. Similarly, these objects are classified as 'aligned' if the value in the aligned slot is TRUE. Thus, align will only operate on an 'unaligned' wavelet transform object if inverse = FALSE and on an 'aligned' wavelet transform object if inverse = TRUE.

The argument coe is passed to the wt.filter.shift function to determine what method to use for computing phase shifts (see documentation for wt.filter.shift).

## Value

Either an aligned or unaligned object of the same class as wt (see Details above).

## Author(s)

Eric Aldrich. ealdrich@gmail.com.

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[wt.filter.shift](#page-39-1), [dwt](#page-2-1), [modwt](#page-15-1).

## Examples

```
# obtain the two series listed in Percival and Walden (2000), page 42
X1 \leftarrow c(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2, .1,-.1,.1,.7,.9,0,.3)X2 \leq c \leq C(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2,1,-.1,1,-.7,0,0,0,3)
```

```
# combine them and compute DWT
newX \leftarrow \text{cbind}(X1, X2)wt <- dwt(newX, n.levels=3, boundary="reflection", fast=FALSE)
# align
```

```
wt.aligned <- align(wt)
```
<span id="page-2-1"></span>

dwt *Discrete Wavelet Transform*

## Description

Computes the discrete wavelet transform coefficients for a univariate or multivariate time series.

#### Usage

```
dwt(X, filter="la8", n.levels, boundary="periodic", fast=TRUE)
```


## Details

The discrete wavelet transform is computed via the pyramid algorithm, using pseudocode written by Percival and Walden (2000), pp. 100-101. When boundary="periodic" the resulting wavelet and scaling coefficients are computed without making changes to the original series - the pyramid algorithm treats X as if it is circular. However, when boundary="reflection" a call is made to extend. series, resulting in a new series which is reflected to twice the length of the original series. The wavelet and scaling coefficients are then computed by using a periodic boundary condition on the reflected sereis, resulting in twice as many wavelet and scaling coefficients at each level.

## Value

Returns an object of class dwt, which is an S4 object with slots



#### <span id="page-4-0"></span>dwt-class 5



## Author(s)

Eric Aldrich. ealdrich@gmail.com.

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[modwt](#page-15-1), [wt.filter](#page-34-1).

## Examples

# obtain the two series listed in Percival and Walden (2000), page 42  $X1 \leq C \leq C \leq 2, -14, -16, -15, -18, -14, -19, 0, -12, 1, -11, 11, 17, 19, 0, 13$  $X2 \leq -c(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2,1,-.1,1,-.7,0,0,0,3)$ 

```
# combine them and compute DWT
newX \leftarrow \text{cbind}(X1, X2)wt <- dwt(newX, n.levels=3, boundary="reflection", fast=FALSE)
```
<span id="page-4-1"></span>dwt-class *Discrete Wavelet Transform Object*

## Description

An S4 object containing discrete wavelet transform information.

#### <span id="page-5-0"></span>**Slots**

- W A list with element  $i$  comprised of a matrix containing the  $i$ th level wavelet coefficients.
- $V$  A list with element i comprised of a matrix containing the ith level scaling coefficients.
- filter A wt.filter object containing information for the filter used in the decomposition. See help(wt.filter) for details.
- level An integer value representing the level of wavelet decomposition.
- **n.boundary** A numeric vector indicating the number of boundary coefficients at each level of the decomposition.
- boundary A character string indicating the boundary method used in the decomposition. Valid values are "periodic" or "reflection".
- series The original time series, X, in matrix format.
- class.X A character string indicating the class of the input series. Possible values are "ts", "mts", "numeric", "matrix", or "data.frame".
- attr. $X$  A list containing the attributes information of the original time series, X. This is useful if  $X$  is an object of class ts or mts and it is desired to retain relevant time information. If the original time series, X, is a matrix or has no attributes, then attr.X is an empty list.
- aligned A logical value indicating whether the wavelet and scaling coefficients have been phase shifted so as to be aligned with relevant time information from the original series. The value of this slot is initially FALSE and can only be changed to TRUE via the align function, with the dwt object as input.
- coe A logical value indicating whether the center of energy method was used in phase alignement of the wavelet and scaling coefficients. By default, this value is FALSE (and will always be FALSE when aligned is FALSE) and will be set to true if the dwt object is phase shifted via the align function and center of energy method.

## Author(s)

Eric Aldrich. ealdrich@gmail.com.

#### References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[dwt](#page-2-1), [modwt](#page-15-1), [modwt-class](#page-17-1), [wt.filter](#page-34-1).

<span id="page-6-0"></span>

## Description

Implementation of DWT and MODWT forward and backward pyramid algorithms.

## Usage

```
dwt.forward(V, filter)
dwt.backward(W, V, filter)
modwt.forward(V, filter, j)
modwt.backward(W, V, filter, j)
```
#### Arguments



## Details

An implementation of the DWT and MODWT forward and backward pyramid algorithms using pseudocode written by Percival and Walden (2000), pp. 100-101, 177-178. These functions are intended primarily as helper functions for the dwt, modwt, idwt and imodwt functions.

## Value

dwt.forward and modwt.forward return a list of two elements containing vectors of wavelet and scaling coefficients for the subsequent level of analysis. dwt.backward and modwt.backward return a vector of scaling coefficients for the previous level of analysis.

## Author(s)

Eric Aldrich. ealdrich@gmail.com.

### References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[dwt](#page-2-1), [modwt](#page-15-1), [wt.filter](#page-34-1).

#### Examples

```
# obtain the two series listed in Percival and Walden (2000), page 42
X1 \leftarrow c(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2,.1,-.1,.1,.7,.9,0,.3)X2 \leq -c(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2, .1,-.1, .1,-.7, .9,0,.3)# compute the LA8 wavelet filter for both DWT and MODWT
la8.dwt <- wt.filter()
la8.modwt <- wt.filter(modwt=TRUE)
# compute the DWT and MODWT level one wavelet and scaling coefficients
wt.dwt <- dwt.forward(X1, la8.dwt)
wt.modwt <- modwt.forward(X2, la8.modwt, 1)
# compute the original series with the level one coefficients
newX.dwt <- dwt.backward(wt.dwt$W, wt.dwt$V, la8.dwt)
newX.modwt <- modwt.backward(wt.modwt$W, wt.modwt$V, la8.modwt, 1)
```
ecg *Electrocardiogram Data*

#### Description

Electrocardiogram measurements for an individual with arhythmia. Observations were made over an 11.37 second interval at a rate of 180 samples per second. For more detailed information regarding the collection of this data, see Percival and Walden (2000), p. 125.

#### Usage

data(ecg)

#### Format

A time series object containing 2048 observations.

## Source

http://www.ms.washington.edu/~s530/data.html

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press, sec. 4.10.

<span id="page-7-0"></span> $8$  ecg

<span id="page-8-0"></span>extend.series *Extend a Time Series*

#### **Description**

Extends a univariate or multivariate time series beyond its original length.

#### Usage

```
extend.series(X, method="reflection", length="double", n, j)
```
## Arguments



## Details

The original time series, X, is extended to a new length as determined by length: if length = "arbitrary", a value for the argument n must be specified in order to extend the series to length n; if length = "power of 2" the series will be extended to the nearst multiple of  $2^j$ , where j must be specified as an argument to the function; if length = "double" the series will be extended to twice its original length.

Once the length of the extended series is determined, the actual extension is performed using one of five methods: if method = "periodic" a periodic extension is made by concatenating the series with itself (without changing the order of the values); if method = "reflection" an extension is made by concatenating the series with a reflection (reverse ordering) of itself; if method = "zero" the series is padded with zeros; if method = "mean" the series is padded with its mean; if method = "reflection.inverse" an extension is made by concatenating the series with a reflection (reverse ordering) of itself that is further reflected of the horizontal axis  $y = x_N$  where  $x_N$  is the last value of the observed series.

In the case that the series is extended beyond twice it's original length, the "periodic" and "reflection" methods are repetetively applied so that every subseries of length  $N$  (the length of the original series) is either a periodic or reflected extension of any other  $N$  adjacent values. Similarly, the "reflection.inverse" method is repetetively applied so that every  $2N$  values are a periodic extension of any adjacent  $2N$  values.

## <span id="page-9-0"></span>Value

The extended time series, in its original class.

## Author(s)

Eric Aldrich. ealdrich@gmail.com.

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

#### See Also

[dwt](#page-2-1), [modwt](#page-15-1).

#### Examples

# obtain the two series listed in Percival and Walden (2000), page 42  $X1 \leftarrow c(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2,.1,-.1,.1,.7,.9,0,.3)$  $X2 \leq -c(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2, .1,-.1, .1,-.7, .9,0,.3)$ 

```
# combine them and extend using reflection method
newX < - \text{cbind}(X1, X2)newX.ext1 <- extend.series(newX, length="arbitrary", n=64)
plot.ts(newX.ext1)
# apply another extension method
```

```
newX.ext2 <- extend.series(newX, method="reflection.inverse",
                           length="powerof2", j=6)
plot.ts(newX.ext2)
```
figure108.wt.filter *Plot Multiple DWT Wavelet or Scaling Filters*

## Description

Plots multiple DWT Wavelet or Scaling Filters similar to Figure 108 in *Wavelet Methods for Time Series Analysis* by Percival and Walden (2000).

## Usage

```
figure108.wt.filter(filter.objects, level = 1, l = NULL, wavelet = TRUE)
```
## <span id="page-10-0"></span>**Arguments**



## Details

The plotting space available for each filter is dictated by the value of greatest magnitude of all the filters plotted. The vertical plotting space for each level will then be 2 times the absolute value of this magnitude.

The filters are successively plotted in the order given in filter.object, where the first filter in filter.object is drawn at the top of the plot region, and the successive filters are plotted below.

## Author(s)

Kelvin Ma, kkym@u.washington.edu

#### References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[wt.filter](#page-34-1)

## Examples

```
# Plotting the LA8 Wavelet Filter
filter <- wt.filter()
figure108.wt.filter(filter)
```

```
# Alternatively
figure108.wt.filter("la8")
```
# Plotting the Haar, D4, D6 Wavelet Filters figure108.wt.filter(list("haar", "d4", "d6"))

```
# Plotting the Haar, D4, D6 Scaling Filters
figure108.wt.filter(list("haar", "d4", "d6"), wavelet = FALSE)
```

```
# Alternatively
haar <- wt.filter("haar")
d6 <- wt.filter("d6")
figure108.wt.filter(list(haar, "d4", d6), wavelet = FALSE)
# Adding an "made up" filter (represented by c(1,-1,1,-1)
figure108.wt.filter(list(haar, "d4", d6, c(1,-1,1,-1)), wavelet = FALSE)
```
figure98.wt.filter *Plot a DWT Wavelet or Scaling Filter for Specific Levels.*

#### Description

Plots a DWT Wavelet or Scaling Filter for specific levels similar to Figure 98 in *Wavelet Methods for Time Series Analysis* by Percival and Walden (2000).

## Usage

```
figure98.wt.filter(filter, levels = NULL, wavelet = TRUE, y.normalize =
TRUE)
```
## Arguments



#### Details

If a single number is specified for levels, then the filter of levels 1 through levels will be plotted. Otherwise, a vector will specify which levels of the wavelet coefficients will be plotted. If levels is not defined, it will default to the vector 1:7.

The plotting space available for each level of the filter is dictated by the value of greatest magnitude of all the levels of the filter plotted. The vertical plotting space for each level will then be 2 times the absolute value of this magnitude.

The label 'L sub j' on the horizontal axis varies from level to level, but 'L sub j' indicates the length of a filter at level j. The filter vector at a given level j is indexed from 0 to 'L sub j' - 1. 'L sub j' is defined by formula 96a in *Wavelet Methods for Time Series Analysis* by Percival and Walden (2000).

<span id="page-11-0"></span>

<span id="page-12-0"></span>idwt 13

## Author(s)

Kelvin Ma, kkym@u.washington.edu

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[wt.filter](#page-34-1)

## Examples

```
# Plotting LA8 Wavelet Filter Coefficients Levels 1 through 7.
filter <- wt.filter()
figure98.wt.filter(filter)
# Alternatively
figure98.wt.filter("la8")
# Plotting D4 Scaling Filter Coefficients Levels 1, 3, and 5 and not
# vertically normalizing each level to its plotting region.
figure98.wt.filter("d4", levels = c(1,3,5), wavelet = FALSE, y.normalize
= FALSE)
```
<span id="page-12-1"></span>idwt *Inverse Discrete Wavelet Transform*

## Description

Computes the inverse discrete wavelet transform for a discrete wavelet transform that was obtained from a univariate or multivariate time series.

## Usage

idwt(wt, fast=TRUE)

#### **Arguments**



## Details

The inverse discrete wavelet transform is computed via the inverse pyramid algorithm, using pseudocode written by Percival and Walden (2000), p. 101.

## <span id="page-13-0"></span>Value

An object with class and attributes equivalent to the original series that was used to compute the DWT. In general, the output will be equivalent to the original series (i.e.  $X = \text{idwt}(\text{dwt}(X))$ ), however when thresholding or shrinkage methods are used on the dwt object, the output of idwt may differ from the original series.

## Author(s)

Eric Aldrich. ealdrich@gmail.com.

#### References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[dwt](#page-2-1), [modwt](#page-15-1), [imodwt](#page-13-1).

## Examples

```
# obtain the two series listed in Percival and Walden (2000), page 42
X1 \leftarrow c(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2,1,-.1,1,1,7,9,0,.3)X2 \leq c \leq C(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2,.1,-.1,.1,-.7,.9,0,.3)
```

```
# combine them and compute DWT
newX \leftarrow \text{cbind}(X1, X2)wt <- dwt(newX, n.levels=3, boundary="reflection")
```

```
# compute the inverse DWT
invX \leftarrow idwt(wt)
```
# compare newX invX

<span id="page-13-1"></span>imodwt *Inverse Maximal Overlap Discrete Wavelet Transform*

## Description

Computes the inverse maximal overlap discrete wavelet transform for a maximal overlap discrete wavelet transform that was obtained from a univariate or multivariate time series.

#### Usage

imodwt(wt, fast=TRUE)

#### <span id="page-14-0"></span>imodwt the contract of the con

#### Arguments



## Details

The inverse discrete wavelet transform is computed via the inverse pyramid algorithm, using pseudocode written by Percival and Walden (2000), p. 101.

## Value

An object with class and attributes equivalent to the original series that was used to compute the MODWT. In general, the output will be equivalent to the original series (i.e.  $X = \text{imodwt}(\text{modwt}(X)))$ , however when thresholding or shrinkage methods are used on the modwt object, the output of imodwt may differ from the original series.

## Author(s)

Eric Aldrich. ealdrich@gmail.com.

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[dwt](#page-2-1), [modwt](#page-15-1), [idwt](#page-12-1).

#### Examples

```
# obtain the two series listed in Percival and Walden (2000), page 42
X1 \leftarrow c(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2,.1,-.1,.1,.7,.9,0,.3)X2 \leq -c(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2, .1,-.1, .1,-.7, .9,0,.3)
```

```
# combine them and compute MODWT
newX \leftarrow \text{cbind}(X1, X2)wt <- modwt(newX, n.levels=3, boundary="reflection")
```

```
# compute the inverse MODWT
invX \leftarrow imodwt(wt)
```
# compare newX invX

## <span id="page-15-1"></span><span id="page-15-0"></span>Description

Computes the maximal overlap discrete wavelet transform coefficients for a univariate or multivariate time series.

## Usage

```
modwt(X, filter="la8", n.levels, boundary="periodic", fast=TRUE)
```
## Arguments



## Details

The maximal overlap discrete wavelet transform is computed via the pyramid algorithm, using pseudocode written by Percival and Walden (2000), p. 178. When boundary="periodic" the resulting wavelet and scaling coefficients are computed without making changes to the original series - the pyramid algorithm treats X as if it is circular. However, when boundary="reflection" a call is made to extend. series, resulting in a new series which is reflected to twice the length of the original series. The wavelet and scaling coefficients are then computed by using a periodic boundary condition on the reflected sereis, resulting in twice as many wavelet and scaling coefficients at each level.

#### Value

Returns an object of class modwt, which is an S4 object with slots

W A list with element i comprised of a matrix containing the ith level wavelet coefficients.

<span id="page-16-0"></span>modwt the contract of the cont



## Author(s)

Eric Aldrich. ealdrich@gmail.com.

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[dwt](#page-2-1), [wt.filter](#page-34-1).

## Examples

# obtain the two series listed in Percival and Walden (2000), page 42  $X1 \leftarrow c(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2,.1,-.1,.1,.7,.9,0,.3)$  $X2 \leftarrow c(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2,.1,-.1,.1,-.7,.9,0,.3)$ 

# combine them and compute DWT newX <- cbind(X1,X2) wt <- dwt(newX, n.level=3, boundary="reflection", fast=FALSE)

<span id="page-17-1"></span><span id="page-17-0"></span>

## **Description**

An S4 object containing maximal overlap discrete wavelet transform information.

## **Slots**

W A list with element  $i$  comprised of a matrix containing the  $i$ th level wavelet coefficients.

- V A list with element i comprised of a matrix containing the *i*th level scaling coefficients.
- filter A wt.filter object containing information for the filter used in the decomposition. See help(wt.filter) for details.

level An integer value representing the level of wavelet decomposition.

- n.boundary A numeric vector indicating the number of boundary coefficients at each level of the decomposition.
- boundary A character string indicating the boundary method used in the decomposition. Valid values are "periodic" or "reflection".
- series The original time series, X, in matrix format.
- class.X A character string indicating the class of the input series. Possible values are "ts", "mts", "numeric", "matrix", or "data.frame".
- **attr.X** A list containing the attributes information of the original time series, X. This is useful if  $X$  is an object of class ts or mts and it is desired to retain relevant time information. If the original time series, X, is a matrix or has no attributes, then attr.X is an empty list.
- aligned A logical value indicating whether the wavelet and scaling coefficients have been phase shifted so as to be aligned with relevant time information from the original series. The value of this slot is initially FALSE and can only be changed to TRUE via the align function, with the modwt object as input.
- coe A logical value indicating whether the center of energy method was used in phase alignement of the wavelet and scaling coefficients. By default, this value is FALSE (and will always be FALSE when aligned is FALSE) and will be set to true if the modwt object is phase shifted via the align function and center of energy method.

#### Author(s)

Eric Aldrich. ealdrich@gmail.com.

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[dwt](#page-2-1), [dwt-class](#page-4-1), [modwt](#page-15-1), [wt.filter](#page-34-1).

## Description

Computes the multiresolution analysis for a univariate or multivariate time series.

## Usage

```
mra(X, filter="la8", n.levels, boundary="periodic", fast=TRUE, method="dwt")
```
## Arguments



## Value

Returns an object of class mra, which is an S4 object with slots



<span id="page-18-0"></span>

<span id="page-19-0"></span>

## Author(s)

Eric Aldrich. ealdrich@gmail.com.

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[dwt](#page-2-1), [modwt](#page-15-1), [wt.filter](#page-34-1).

## Examples

# obtain the two series listed in Percival and Walden (2000), page 42  $X1 \leftarrow c(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2, .1,-.1,.1,.7,.9,0,.3)$  $X2 \leq c \leq C(.2,-.4,-.6,-.5,-.8,-.4,-.9,0,-.2,1,-.1,1,-.7,0,0,0.3)$ 

# combine them and compute MRA  $newX \leftarrow \text{cbind}(X1, X2)$ mra.out <- mra(newX, n.levels=3, boundary="reflection")

mra-class *Multiresolution Analysis Object*

#### Description

An S4 object containing multiresolution analysis information.

## **Slots**

 **A list with element i comprised of a matrix containing the** *i***th level wavelet detail.** 

 $S$  A list with element i comprised of a matrix containing the ith level wavelet smooths.

filter A wt.filter object containing information for the filter used in the decomposition. See help(wt.filter) for details.

level An integer value representing the level of wavelet decomposition.

- <span id="page-20-0"></span>**boundary** A character string indicating the boundary method used in the wavelet decomposition. Valid values are "periodic" or "reflection".
- series The original time series,  $X$ , in matrix format.
- class.X A character string indicating the class of the input series. Possible values are "ts", "mts", "numeric", "matrix", or "data.frame".
- **attr.X** A list containing the attributes information of the original time series, X. This is useful if  $X$  is an object of class ts or mts and it is desired to retain relevant time information. If the original time series, X, is a matrix or has no attributes, then attr.X is an empty list.
- method A character string indicating which type of wavelet decomposition was performed (either "dwt" or "modwt").

#### Author(s)

Eric Aldrich. ealdrich@gmail.com.

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[dwt](#page-2-1), [modwt](#page-15-1), [wt.filter](#page-34-1).

nile *Yearly Nile River Minima*

#### Description

Yearly minimum water level of the Nile River starting in 622 A.D. and ending in 1284 A.D. For more information see Percival and Walden (2000), p. 190.

#### Usage

data(nile)

## Format

A time series object containing 683 observations.

## Source

http://www.ms.washington.edu/~s530/data.html

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press, sec. 5.9.

<span id="page-21-0"></span>

## Description

Measurements of vertical ocean sheer, obtained by dropping an instrument veritcally into the ocean and recording observations at 0.1 meter intervals. Starting depth for the series is 350.0 meters and ending depth is 1037.4 meters. For more detailed information regarding the collection of this data, see Percival and Walden (2000), p. 193.

#### Usage

data(ocean)

## Format

A time series object containing 6875 observations.

#### Source

http://www.ms.washington.edu/~s530/data.html

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press, sec. 5.10.

<span id="page-21-1"></span>plot.dwt *Plot DWT Coefficients*

#### Description

Plot DWT wavelet and scaling coefficients.

## Usage

```
## S3 method for class 'dwt'
plot(x, levels = NULL, draw.boundary = FALSE, type = "stack",col.plot = "black", col.boundary = "red", X.xtick.at = NULL, X.ytick.at
= NULL, Stack.xtick.at = NULL, Stack.ytick.at = NULL, X.xlab = "t",
y.rlabs = TRUE, plot.X = TRUE, plot.W = TRUE, plot.V = TRUE, \ldots)
```
#### plot.dwt 23

## Arguments



#### Details

plot.dwt plots the DWT wavelet and scaling coefficients. The type parameter species the type of plot, which is currently only stack. If the DWT object is defined for multiple time series, only the data pertaining to the first time series of the DWT object is plotted. Thus, only the wavelet coefficients and scaling coefficients of the first time series of the DWT object will be plotted.

If a single number is specified for levels, then the wavelet coefficients of levels 1 through levels will be plotted. Otherwise, a vector or the first element of a list will specify which levels of the wavelet coefficients will be plotted. Unless specified in the second element of a list, only one level of scaling coefficients will be plotted and this level is equal to the highest level of the wavelet coefficients plotted.

<span id="page-23-0"></span>For each respective axis, the distance between a tick mark on the plot of the original time series is equivalent to the distance between a tick mark on the stacked plot. Thus, when altering the relative spacing of the tick marks on the plot of the original time series using X.xtick.at or X.ytick.at, the tick marks of the stacked plot are automatically adjusted. If the plot of the original time series is not drawn, then the user can alter the spacing of the tick marks using Stack.xtick.at and Stack.ytick.at for the horizontal and vertical axes, respectively.

One of plot.W or plot.V must be TRUE.

## Author(s)

Kelvin Ma, kkym@u.washington.edu

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[stackplot](#page-30-1)

## Examples

```
X < - rnorm(2048)
dwtobj \leftarrow dwt(X)# Plotting wavelet coefficients of levels 1 through 6 and scaling
# coefficients of level 6.
plot.dwt(dwtobj, levels = 6)
# Plotting wavelet coefficients of levels 1, 3, 5, and scaling
# coefficients of levels 4 and 5, and green boundary coefficient lines.
plot.dwt(dwtobj, levels = list(c(1,3,5),c(4,5)), draw.boundary = TRUE,
col.boundary = "green")
# Plotting wavelet coefficients of level 1 through 6 and not plotting
# any scaling coefficients.
plot.dwt(dwtobj, levels = 6, plot.V = FALSE)
```
plot.dwt.multiple *Plot Multiple DWT Objects*

## Description

Plot wavelet and scaling coefficients of multiple DWT objects.

## plot.dwt.multiple 25

## Usage

```
## S3 method for class 'multiple'
plot.dwt(x, levels = NULL, ylim = NULL, draw.dashed.lines =TRUE, draw.level.labels = TRUE, col = c("red", "blue"), ...
```
#### Arguments



#### Details

If a single number is specified for levels, then the wavelet coefficients of levels 1 through levels will be plotted. Otherwise, a vector or the first element of a list will specify which levels of the wavelet coefficients will be plotted. Unless specified in the second element of a list, only one level of scaling coefficients will be plotted and this level is equal to the highest level of the wavelet coefficients plotted. If a DWT object is defined for multiple time series, only the data pertaining to the first time series of the DWT object is plotted. Thus, only the wavelet coefficients and scaling coefficients of the first time series of the DWT objects will be plotted.

For each dwtobject in the list of x, plot.dwt.multiple takes the coefficients of the dwt object and concatenates wavelet coefficients and scaling coefficients by levels specified in levels. The wavelet coefficients will always be plotted preceding the scaling coefficients.

This function allows users to visually examine differences in the DWT transform of a time series using different filters (different dwt objects).

For an example, see Figure 126 of *Wavelet Methods for Time Series Analysis* by Percival and Walden (2000).

#### Author(s)

Kelvin Ma, kkym@u.washington.edu

#### References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[plot.dwt](#page-21-1)

## Examples

```
X < - rnorm(2048)# Create DWT Object of X with the "la8" filter.
dwtobj1 \leftarrow \text{dwt}(X, \text{ filter} = "la8")# Create DWT Object of X with the "d4" filter.
dwtobj2 <- dwt(X, filter = "d4")
# Create DWT Object of X with the "haar" filter
dwtobj3 \leq - dwt(X, filter = "haar")
# Create DWT Object of X with the "c6" filter
dwtobj4 <- dwt(X, filter = "c6")
#Create list of dwt objects
dwtlist <- list(dwtobj1, dwtobj2, dwtobj3, dwtobj4)
# Plot the dwt objects and the wavelet coefficients of level 1 through 6
# and the scaling coefficients of level 6. The first level drawn will
# be purple and the next level drawn will be gold.
plot.dwt.multiple(dwtlist, levels = 6, col = c("purple", "gold"))
# Plot the dwt objects and the wavelet coefficients of level 1, 3, and 5
# and scaling coefficients of level 2, and 4.
plot.dwt.multiple(dwtlist, levels = list(c(1,3,5), c(2,4)))
```
<span id="page-25-1"></span>

plot.modwt *Plot MODWT Coefficients*

## **Description**

Plot MODWT wavelet and scaling coefficients.

## Usage

```
## S3 method for class 'modwt'
plot(x, levels = NULL, draw.boundary = FALSE, type = "stack",col.plot = "black", col.boundary = "red", X.xtick.at = NULL, X.ytick.at
= NULL, Stack.xtick.at = NULL, Stack.ytick.at = NULL, X.xlab = "t",
y.rlabs = TRUE, plot.X = TRUE, plot.W = TRUE, plot.V = TRUE, ...)
```
#### Arguments



<span id="page-25-0"></span>



## Details

plot.modwt plots the MODWT wavelet and scaling coefficients. The type parameter species the type of plot, which is currently only stack. If the MODWT object is defined for multiple time series, only the data pertaining to the first time series of the MODWT object is plotted. Thus, only the wavelet coefficients and scaling coefficients of the first time series of the MODWT object will be plotted. If a single number is specified for levels, then the wavelet coefficients of levels 1 through levels will be plotted. Otherwise, a vector or the first element of a list will specify which levels of the wavelet coefficients will be plotted. Unless specified in the second element of a list, only one level of scaling coefficients will be plotted and this level is equal to the highest level of the wavelet coefficients plotted.

For each respective axis, the distance between a tick mark on the plot of the original time series is equivalent to the distance between a tick mark on the stacked plot. Thus, when altering the relative spacing of the tick marks on the plot of the original time series using X.xtick.at or X.ytick.at, the tick marks of the stacked plot are automatically adjusted. If the plot of the original time series <span id="page-27-0"></span>is not drawn, then the user can alter the spacing of the tick marks using Stack.xtick.at and Stack.ytick.at for the horizontal and vertical axes, respectively.

One of plot.W or plot.V must be TRUE.

## Author(s)

Kelvin Ma, kkym@u.washington.edu

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

#### See Also

[stackplot](#page-30-1)

## Examples

```
X \leftarrow \text{norm}(2048)modw \t{tobj} \leq m \t{right}(\t{X})# Plotting wavelet coefficients at levels 1 through 6 and scaling
# coefficients at level 6.
plot.modwt(modwtobj, levels = 6)
# Plotting wavelet coefficients of levels 1, 3, 5, and scaling
# coefficients of levels 4 and 5, and green boundary coefficients
# lines.
plot.modwt(modwtobj, levels = list(c(1,3,5),c(4,5)), draw.boundary =
TRUE, col.boundary = "green")
# Plotting wavelet coefficients at level 1 through 6 and not plotting
# any scaling coefficients.
plot.modwt(modwtobj, levels = 6, plot.V = FALSE)
```
<span id="page-27-1"></span>scalingshift.dwt *Find Circular Shift of DWT Scaling Coefficients*

#### Description

Finds circular shift of DWT scaling coefficients at a specified level, for a specified filter length.

## Usage

scalingshift.dwt(L, j, N = NULL)

## <span id="page-28-0"></span>scalingshift.dwt 29

#### **Arguments**



## Details

This function computes the circular shift associated with a wavelet transform filter of length L and level j. L must be of even length.

Each coefficient's index will be circularly shifted forward by the value outputted. For example, if we have a vector of DWT scaling coefficients (with NAs inserted) with indices 1,2,3,4,5,6,7,8 and the value outputted from the function scalingshift.dwt is 2, the new order of the vector of aligned DWT wavelet coefficients would be 7,8,1,2,3,4,5,6.

If N is specified, the function will output the "shift" Modulo N. Because the shift is circular, the value outputted when N is specified is equivalent to the case when N is not specified.

This function calculates the absolute value of vjH in formulas (114c) in *Wavelet Methods for Time Series Analysis* by Percival and Walden (2000). (And optionally vjH Modulo N)

## Value



## Author(s)

Kelvin Ma, kkym@u.washington.edu

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

#### See Also

[waveletshift.dwt](#page-33-1) and [plot.dwt](#page-21-1)

## Examples

```
# Finding the circular shift for a wavelet transform filter of length 8,
# pertaining to DWT Scaling Coefficients of level 5.
scalingshift.dwt(8, 5)
```

```
# If the sample size of the original time series is of length 1024.
scalingshift.dwt(8, 5, N = 1024)
```
<span id="page-29-0"></span>squaredgain.wt.filter *Plot Squared Gain Function for Wavelet Filter*

## Description

Plots squared gain function for a wavelet filter.

## Usage

```
squaredgain.wt.filter(filter, level = 1, N = NULL, draw.bands = TRUE,
wavelet = TRUE)
```
## Arguments



## Details

If N is not specified, and the filter supplied is of length less than 1024, then the filter vector will be padded to a length of 1024. Otherwise, if the filter supplied is of length greater than 1024, then the filter vector will be padded to the first 'power of 2' that is greater than the length of the filter supplied.

## Author(s)

Kelvin Ma, kkym@u.washington.edu

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[wt.filter](#page-34-1)

#### <span id="page-30-0"></span>stackplot 31

## Examples

```
# Plotting the squared gain function of LA8 Wavelet Filter Coefficients.
filter <- wt.filter("la8")
squaredgain.wt.filter(filter)
# Plotting the squared gain function of LA8 Scaling Filter Coefficients.
squaredgain.wt.filter(filter, wavelet = FALSE)
# Plotting the squared gain function of Haar Wavelet Filter Coefficients
# without supplying a filter object.
squaredgain.wt.filter("haar")
```
<span id="page-30-1"></span>stackplot *Stacked Plotting*

## Description

Plot an object as a stacked set of plots, plotting each column using the same x-axis and plotting the y-values at the same scale with an offset equal to the magnitude of each column vector.

## Usage

```
stackplot(x, y = NULL, type = "h", axes.labels = FALSE,
xlab = "", ylab = "", yllabs = NULL, yrlabs = NULL, draw.divides =TRUE, xtick.at = NULL, ytick.at = NULL, col = "black", main = "")
```
## Arguments



<span id="page-31-0"></span>

## Details

Each object specified by x or y will be coerced into a list, and each object of the list will be coerced into a matrix using the function as.matrix. Each column of a matrix will represent the vertical y-values for each individual plot in the stacked plot. Horizontal x-values are discussed below.

stackplot will plot a matrix of values as a series of stacked plots, where the individual columns are plotted using the same x and y scales, but offset by the magnitude  $(= abs(max-min))$  of the column.

If the object to be plotted is a Time Series object, the start and end values of the object will represent the first and last horizontal x-values of the points to be plotted. The Time Series will be coerced into matrix.

The limits of the horizontal x-axis and vertical y-axis will be dictated by the data in the first object of the list.

If the object to be plotted is a list of matrices, and  $x$  is not specified, the horizontal  $x$ -values of the points to be plotted for a given matrix will default to the rownames of that matrix. The rownames can be either numeric or character. If the first matrix of the list does not have rownames, then the horizontal x-values for each matrix (and its corresponding plot) will default to range from 1 to the length of the columns of the respective matrix.

If the first matrix of the list does have specified rownames, then the horizontal x-values assume the values of the rownames of the respective matrix. However, if another matrix of the list does not have specified rownames, then its horizontal x-values will default to range from 1 to the length of the columns of that respective matrix.

The user can specify the type of plot (currently only "h" and "l") to be drawn for every element of the list in x or y using the vector type. If the Nth element  $(N>1)$  in type is NA and the Nth element of the list in x or y exists, then the type of plot for the Nth element of the list will be equal to the type of plot for the N-1th element of the list. Similarly for col, which specifies the colors for each element of the list in x or y.

The labels y. 11abs and y. rlabs will be placed at the "zero" lines of each individual plot starting from the bottom working up to the top.

## Author(s)

Kelvin Ma, kkym@u.washington.edu

## See Also

[plot.dwt](#page-21-1) and [plot.modwt](#page-25-1)

#### <span id="page-32-0"></span>subtidal 33

## Examples

```
A <- array(c(1,2,3,4), c(3,4))# Plotting the vectors c(1,2,3), c(4,1,2), c(3,4,1), and c(2,3,4) in a
# stacked plot with 4 separate plots). The x-values will be 1:3.
stackplot(A)
#Plotting A with numeric labels on the axis and making the plot "red".
#Also labeling the x-axis with "X-Label", and the y-axis with "Y-Label".
stackplot(A, axes.labels = TRUE, xlab = "X-Label", ylab = "Y-Label")
timeSeries <- ts(A)
# Plotting the Time Series created by coercing the array A into a Time
# Series. Observe that this plot similar to the plot in the first example.
stackplot(timeSeries)
lfig <- 2:11
hfig <-1:10# Plotting a stacked plot of two types: histogram and lines. Where the
# line joins the points (1,2) and (10,11), and the histogram is similar to
# the histogram formed by entering plot(1:10, type = "h"). The line will
# be red and the histogram will be yellow.
```

```
stackplot(list(lfig, hfig), type = c("l", "h"), col = c("red", "yellow"))
```
subtidal *Subtidal Sea Level Measurements*

#### **Description**

Wwater level measurements from a stilling well on the open coast in Crescent Bay, CA. Observations were recorded every six minutes, from the beginning of 1980 to the end of 1991. For more detailed information regarding the collection of this data, see Percival and Walden (2000), p. 185.

## Usage

data(subtidal)

## Format

A time series object containing 8746 observations.

## Source

http://www.ms.washington.edu/~s530/data.html

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press, sec. 5.8.

<span id="page-33-1"></span><span id="page-33-0"></span>

## Description

Finds circular shift of DWT wavelet coefficients at a specified level, for a specified filter length.

## Usage

waveletshift.dwt(L, j, N = NULL)

## Arguments



## Details

This function computes the circular shift associated with a wavelet transform filter of length L and level j. L must be of even length.

Each coefficient's index will be circularly shifted forward by the value outputted. For example, if we have a vector of DWT wavelet coefficients (with NAs inserted) with indices 1,2,3,4,5,6,7,8 and the value outputted from the function waveletshift.dwt is 2, the new order of the vector of aligned DWT wavelet coefficients would be 7,8,1,2,3,4,5,6.

If N is specified, the function will output the "shift" Modulo N. Because the shift is circular, the value outputted when N is specified is equivalent to the case when N is not specified.

This function calculates the absolute value of vjG in formulas (114c) in *Wavelet Methods for Time Series Analysis* by Percival and Walden (2000). (And optionally vjG Modulo N)

#### Value

shift Circular shift for specified level of DWT wavelet coefficients for a specified filter length.

## Author(s)

Kelvin Ma, kkym@u.washington.edu

### References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[scalingshift.dwt](#page-27-1) and [plot.dwt](#page-21-1)

#### <span id="page-34-0"></span>wt.filter 35

## Examples

```
# Finding the circular shift for a wavelet transform filter of length 8,
# pertaining to DWT Wavelet Coefficients of level 5.
waveletshift.dwt(8, 5)
# If the sample size of the original time series is of length 1024.
```

```
waveletshift.dwt(8, 5, N = 1024)
```
<span id="page-34-1"></span>wt.filter *Wavelet Transform Filter*

## Description

Generates a wavelet transform filter.

#### Usage

```
wt.filter(filter="la8", modwt=FALSE, level=1)
```
## Arguments



## Details

The character strings currently supported are derived from one of four classes of wavelet transform filters: Daubechies, Least Asymetric, Best Localized and Coiflet. The prefixes for filters of these classes are d, la, bl and c, respectively. Following the prefix, the filter name consists of an integer indicating length. Supported lengths are as follows:

Daubechies 2,4,6,8,10,12,14,16,18,20.

Least Asymetric 8,10,12,14,16,18,20.

```
Best Localized 14,18,20.
```
Coiflet 6,12,18,24,30.

Thus, to obtain the Daubechies wavelet transform filter of length 4, the character string "d4" can be passed to wt.filter.

This naming convention has one exception: the Daubechies wavelet transform filter of length 2 is denoted by haar instead of d2.

## <span id="page-35-0"></span>Value

Returns an object of class wt.filter, which is an S4 object with slots



## Note

The notation h and g for wavelet and scaling coefficients, respectively, follows Percival and Walden (2000). In other texts and articles the reverse notation is often adopted.

## Author(s)

Eric Aldrich. ealdrich@gmail.com.

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[wt.filter.qmf](#page-38-1), [dwt](#page-2-1), [modwt](#page-15-1).

## Examples

```
wt.filter("la14")
```
wt.filter(1:10, modwt=TRUE)

## <span id="page-36-0"></span>**Description**

An S4 object containing wavelet transform filter information.

#### **Slots**

L An integer representing the length of the wavelet and scaling filters.

level An integer representing the level of the wavelet filter.

h A numeric vector of wavelet filter coefficients.

- g A numeric vector of scaling filter coefficients.
- wt.class A character string indicating the class of the wavelet transform filter. Possible values are "Daubechies", "Least Asymetric", "Best Localized", and "Coiflet". If the wt.filter object is generated from a numeric vector of wavelet coefficients, wt.class is "none".
- wt.name A character string indicating the name of the wavlet filter as listed in the Details section, above. If the wt.filter object is generated from a numeric vector of wavelet coefficients, wt.name is "none".
- transform A character string indicating whether the resulting wavelet transform object contains DWT or MODWT coefficients. Possible values are "dwt" and "modwt".

## Note

The notation h and g for wavelet and scaling coefficients, respectively, follows Percival and Walden (2000). In other texts and articles the reverse notation is often adopted.

## Author(s)

Eric Aldrich. ealdrich@gmail.com.

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

#### See Also

[wt.filter.qmf](#page-38-1), [dwt](#page-2-1), [modwt](#page-15-1).

<span id="page-37-0"></span>wt.filter.equivalent *Equivalent Wavelet Transform Filter*

## Description

Generates an equivalent filter for a specified wavelet transform filter and level..

## Usage

```
wt.filter.equivalent(wt.filter, J)
```
## Arguments



#### Details

A wavelet transform of level  $j$  can be computed by iterating the pyramid algorithm  $j$  times with a level 1 wavelet filter, or by simply implementing the pyramid once with the level  $j$  equivalent filter. The equivalent filter is obtained by iteratively filtering the level 1 wavelet filter and scaling filters with upsampled versions of themselves. For details regarding the computation, see Percival and Walden (2000), eqs. 95e and 96e.

## Value

Returns the original wt.filter object with modifications made to the wavelet and scaling filter coefficients, filter length and filter level.

## Author(s)

Eric Aldrich. ealdrich@gmail.com.

#### References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

```
wt.filter.
```
## Examples

```
wf <- wt.filter("la14")
wt.filter.equivalent(wf, 3)
```
<span id="page-38-1"></span><span id="page-38-0"></span>wt.filter.qmf *Quadrature Mirror Filter*

#### Description

Computes the quadrature mirror filter of a series of even length.

## Usage

```
wt.filter.qmf(x, inverse=FALSE)
```
## Arguments



## Details

The quadrature mirror filter is computed as outlined in Percival and Walden, page 75. Specifically, the forward QMF makes use of a wavelet filter to obtain a scaling filter and the inverse QMF does the opposite.

#### Value

A numeric vector which contains the values of x in reverse order, with the values of the resulting vector at even indices being negated in the case that inverse=FALSE and the values at odd indices being negated in the case that inverse=TRUE.

## Author(s)

Eric Aldrich. ealdrich@gmail.com.

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[wt.filter.qmf](#page-38-1), [dwt](#page-2-1), [modwt](#page-15-1).

## Examples

```
# obtain both the la8 wavelet and scaling filters
la8 <- wt.filter("la8")
# now compare output of qmf with values in wt.filter object
wt.filter.qmf(la8@h)
```
<span id="page-39-1"></span><span id="page-39-0"></span>

#### **Description**

Computes phase shifts for wavelet transform coefficients corresponding to a particular filter.

## Usage

wt.filter.shift(filter, J, wavelet=TRUE, coe=FALSE, modwt=FALSE)

#### Arguments



## Details

For wavelet filters of class 'Least Assymetric' or 'Coiflet', the default method for calculating phase shifts is outlined on pages 112-114 and page 124 of Percival and Walden 2000. For wavelet filters of class 'Best Localized', the default shifts are given on page 119 of Percival and Walden 2000. For the Haar filter, both the level  $j$  wavelet and scaling phase shifts are half the length of the level j wavelet and scaling filters and the phase shifts for the  $D(4)$  filter are determined by specifying  $\nu = -1$  in equations (114a) and (114b) of Percival and Walden 2000.

For all other filters, the center of energy method is the default method for computing phase shifts (thus rendering the coe argument irrelevant). If coe = TRUE, then the center of energy method is used regardless of filter class.

By default the DWT phase shifts are computed, using the MODWT phase shifts and the methods outlined on pages 115-116 of Percival and Walden 2000. If modwt = TRUE, then only the MODWT phase shifts are computed.

## Value

A vector of phase shifts, expressed in absolute value. Each element corresponds to the equivalent element in J.

## <span id="page-40-0"></span>wt.filter.shift 41

## Author(s)

Eric Aldrich. ealdrich@gmail.com.

## References

Percival, D. B. and A. T. Walden (2000) *Wavelet Methods for Time Series Analysis*, Cambridge University Press.

## See Also

[wt.filter](#page-34-1), [align](#page-1-1),

## Examples

wt.filter.shift("la14", J=1:6)

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