

Time series topic 1: Weighted regression

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Objectives for the session (2:00 - 3:00)

- What is weighted regression
- The WRTDStidal package
- Application to NERRS data
 - ▶ Fitting a model
 - ► Evaluating a model
 - Viewing a model

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Interactive portion

Follow along as we go:

• flash drive

• online: swmprats.net 2016 workshop tab

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You will run examples whenever you see this guy:





We will use the WRTDStidal package

Option 1, from the R Console prompt:

```
install.packages('WRTDStidal')
library(WRTDStidal)
```



We will use the WRTDStidal package

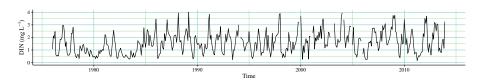
Option 1, from the R Console prompt:

```
install.packages('WRTDStidal')
library(WRTDStidal)
```

Option 2, install the source file from the flash drive:

```
# change as needed
path_to_file <- 'C:/Users/mbeck/Desktop/WRTDStidal_1.0.1.tar.gz'
# install, load
install.packages(path_to_file, repos = NULL, type="source")
library(WRTDStidal)</pre>
```

Observed data represents effects of many processes



Climate

precipitation temperature wind events ENSO effects

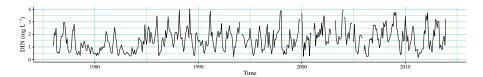
Local

light/turbidity residence time invasive species trophic effects

Regional/historical

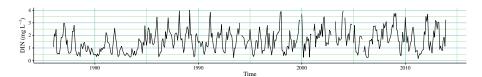
watershed inputs point sources management actions flow changes

Observed data represents effects of many processes

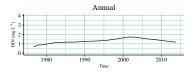


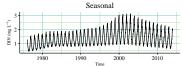
Models should describe components to evaluate effects

Observed data represents effects of many processes

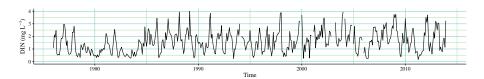


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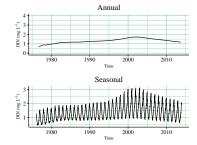


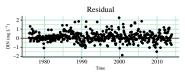


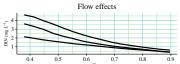
Observed data represents effects of many processes



Models should describe components to evaluate effects







Weighted Regression on Time, Discharge, and Season

- Describes a time series in the context of these parameters, locally fitted
- Useful to describe long-term trends, ie., multi-decadal time series
- Evaluation of flow-normalized trends, hypothesis generation

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Developed by [Hirsch et al., 2010] for pollutants in stream/rivers

Adapted for tidal waters by [Beck and Hagy III, 2015]

How does it work?

$$\ln\left(N\right) = \beta_0 + \beta_1 t + \beta_2 Sal + \beta_3 \sin\left(2\pi t\right) + \beta_4 \cos\left(2\pi t\right)$$

N: nitrogen (or other response endpoint)

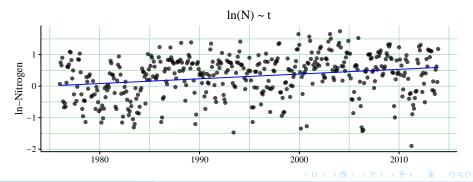
t: time

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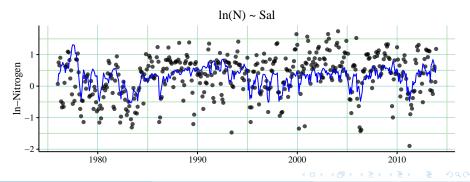


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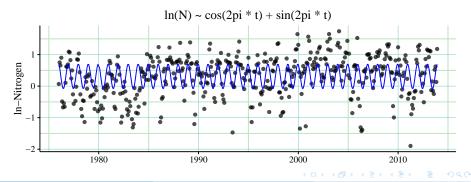


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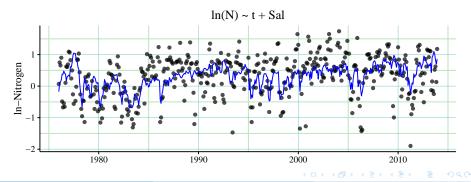


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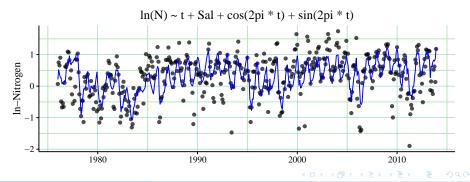


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This is not the whole story...

$$\ln(N) = \beta_0 + \beta_1 t + \beta_2 Sal + \beta_3 \sin(2\pi t) + \beta_4 \cos(2\pi t)$$

One parameter set to many parameter sets - a moving window regression

Within each window, a unique regression is fit, weighted by the local salinity, time, and season

Similar to a loess/spline smooth but specific to the effects of these three variables on the response

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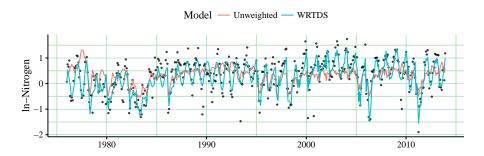
Points: observed time series (black are weighted, grey is zero weight)

Green point: observation at the center of the regression

Blue line: Global model with weights specific to the window

Red line: Accumulated WRTDS model

RMSE fit for unweighted = 0.58, WRTDS = 0.36



All you need to know:

• Describe a response variable in relation to time, salinity (discharge), and season

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...let's not forget about flow-normalization, more about this later

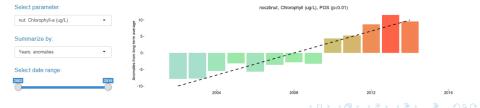
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Trends in SWMP parameters

Created by Marcus W. Beck, beck.marcus@epa.gov, Todd O'Brien, todd.obrien@noaa.gov

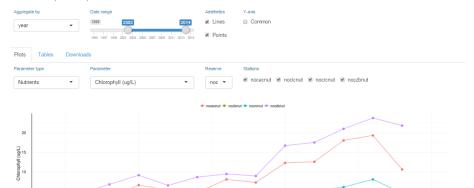
This widept is an interactive tool to explore trends in SNMP data. Trends are described by an increase or decrease in values over time using a simple linear regression of summarized data. The regression for each station are plotted as circles that identify the direction and significance or the tern of the trend direction is blue for decreasing and red for increasing. The significance is indicated by radius of the circle and color shading where larger points with dishers colors indicate a strong trend. Original data are available from they include baruch as cell. See the Gift-but repository for source code. The data include observations through December 2015 (if available) and are current as of May 31, 2015. Please note that the use of simple regression to identify trends is for exploratory purposes only and may not be appropriate for all datasets. The map is centered at 34.04, 47.86 with a zoon level of the simple regression to identify trends is for exploratory purposes only and may not be appropriate for all datasets. The map is centered at 34.04, 47.86 with a zoon level of the simple regression to identify trends is for exploratory purposes only and may not be appropriate for all datasets.



Aggregation of SWMP parameters within/between reserves

Created by Marcus W. Beck, beck.marcus@epa.gov Todd O'Brien, todd.obrien@noaa.gov

This interactive widget can be used to compare time series of site data within and between reserves from the System Wide Monitoring Program of the National Estuarine Research Reserve System (NERRS). Data are based on monthly averages of raw observations through December 2015 and are current as of May 31, 2016. Two plots are shown for selected parameters and reserves that include time series of all sites at each location. The monthly averages are shown by default. Data can also be viewed as quaterly (every timee months) or annual aggregations based on averages of the monthly summaries. Tabular data for each plot can be viewed on the tables tab and downloads of the plots and tables are available on the downloads tab. See the GitHub repository for source ode of to post Issues if foroblems occur.



2010

2015

Using nutrient data from North Carolina NERR, Zeke's Basin site:

1 Import nutrient data, organize

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Using nutrient data from North Carolina NERR, Zeke's Basin site:

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- 6 Assess model performance, plot results

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1 Import nutrient data, organize

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```
# load SWMPr. nutrient data
library(SWMPr)
load(file = 'data/noczbnut.RData')
# rename, qaqc clean up, subset
nut <- noczbnut
nut \leftarrow qaqc(nut, qaqc_keep = c(0, 4))
nut <- subset(nut, select = 'chla n')</pre>
head(nut)
           datetimestamp chla_n
## 1 2002-04-23 15:35:00 2.12
  2 2002-05-24 09:20:00 1.60
  3 2002-06-24 10:35:00 3.47
## 4 2002-07-24 09:40:00 4.43
## 5 2002-08-26 11:31:00 4.65
## 6 2002-09-24 10:40:00 5.95
```



2 Import wq data, organize

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```
# load wa data
load(file = 'data/noczbwq.RData')
# rename, gage clean up, subset
wq <- noczbwq
wq \leftarrow qaqc(wq, qaqc_keep = c(0, 4))
wq <- subset(wq, select = 'sal')</pre>
head(wq)
##
           datetimestamp sal
## 1 2002-03-01 12:30:00 26.8
   2 2002-03-01 13:00:00 26.8
## 3 2002-03-01 13:30:00 26.7
## 4 2002-03-01 14:00:00 26.7
## 5 2002-03-01 14:30:00 26.6
## 6 2002-03-01 15:00:00 26.6
```

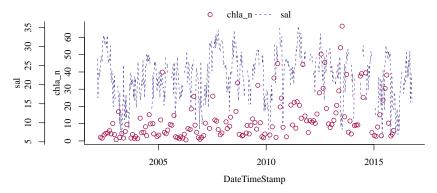


3 Combine chlorophyll and salinity time series

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```
# combine at weekly time step
tomod <- comb(nut, wq, timestep = 'weeks')
# plot both
overplot(tomod, type = c('p', 'l'))</pre>
```



 Prep for WRTDS - requires a tidalmean object with four columns (date, response, salinity/flow, detection limit)

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```
library(WRTDStidal)
# add arbitrary limit column, datetimestamp as date
tomod$lim <- -1e6
tomod$datetimestamp <- as.Date(tomod$datetimestamp)</pre>
# create tidalmean object, note if response is in log or not
tomod <- tidalmean(tomod, reslog = FALSE)</pre>
head(tomod)
##
                   flo lim not cens day num month year dec time
          date res
    2002-02-28 NA 0.7123746 -1e+06
                                                        2 2002 2002 164
                                        NA 0.1643836
  2 2002-03-07 NA 0.6856187 -1e+06 NA 0.1835616 3 2002 2002.184
  3 2002-03-14 NA 0.3712375 -1e+06 NA 0.2027397
                                                       3 2002 2002,203
  4 2002-03-21 NA 0.5685619 -1e+06 NA 0.2219178 3 2002 2002.222
  5 2002-03-28 NA 0.5886288 -1e+06 NA 0.2410959 3 2002 2002.241
                                                        4 2002 2002,260
## 6 2002-04-04 NA 0.6220736 -1e+06
                                        NA 0.2602740
```

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6 Create WRTDS model

5 Create WRTDS model

```
# use modfit function
mod <- modfit(tomod)

##

## Estimating interpolation grid for mean response, % complete...
##

## 5 10 15 20 25 30 35 40 45 50 55 60 65 70 75 80 85 90 95 100
##

## Estimating predictions...
##

##

## Normalizing predictions...</pre>
```

6 Create WRTDS model

```
# use modfit function
mod <- modfit(tomod)

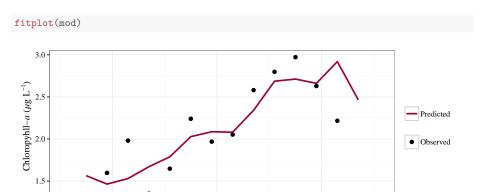
##
## Estimating interpolation grid for mean response, % complete...
##
## 5 10 15 20 25 30 35 40 45 50 55 60 65 70 75 80 85 90 95 100
##
## Estimating predictions...
##
##
## Normalizing predictions...</pre>
```

6 Assess performance

```
## rmse nmse
## 1 0.6417556 0.3476415
```

2005

6 Plot results: fitplot shows observed, predicted, averaged by water years

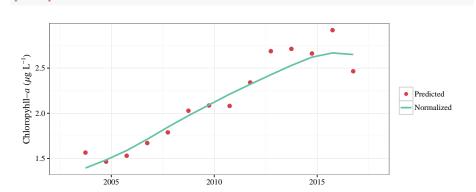


2010

2015

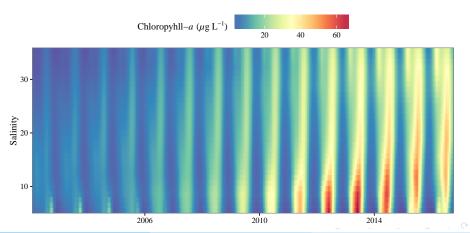
6 Plot results: prdnrmplot shows predicted, flow-normalized predictions, averaged by year

prdnrmplot(mod)



6 Plot results: gridplot shows how the flow, chlorophyll response has changed by season, year

```
gridplot(mod, logspace = F, month = 'all', floscl = F)
```



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Checkout the references and the website for more info: https://github.com/fawda123/WRTDStidal



Up next... Time Series Topic 2: Decomposition

$Questions \ref{eq:questions}$

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References

Beck MW, Hagy III JD. 2015.

Adaptation of a weighted regression approach to evaluate water quality trends in an estuary.

Environmental Modelling and Assessment, 20(6):637-655

Hirsch RM, Moyer DL, Archfield SA. 2010.

Weighted regressions on time, discharge, and season (WRTDS), with an application to Chesapeake Bay river inputs.

Journal of the American Water Resources Association, 46(5):857–880.