Trend Analysis of Censored Metals

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Abstract

This example illustrates the data manipulations for the Tobit analysis of uncensored data. Metals are often left-censored in natural waters and provide a useful example. This example also uses a common time frame for all of the trend tests. The common time frame facilitates comparing trends among the stations. Most often users will want to divide trend analyses into similar groups of analytes like metals, major ions, nutrients, and so forth because they will be analyzed in similar ways and will have common sampling time frames.

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1 Introduction

The data used in this application are a small subset of the data used by Schertz and others (1991). The data are samples taken from water year 1969 (October, 1968) through water year 1989 (September, 1989). Nineteen stations were selected and only copper and iron were selected for the metals.

- > # Load the restrend and other packages and the data
- > library(restrend)
- > library(smwrBase)
- > library(smwrQW)
- > data(EstrendSub)
- > head(EstrendSub)

	STAID	DATI	ES QI	QD	RN.organi	c PN.orga	anic RAm	nonia
1	07227500	1968-10-0	7.6	NA			NA	
2	07227500	1968-10-0	03 5.3	NA			NA	
3	07227500	1968-10-3	16 532.0	NA			NA	
4	07227500	1968-10-3	19 17.0	NA			NA	
5	07227500	1968-11-0	01 17.0	NA			NA	
6	07227500	1968-12-0	01 6.6	NA			NA	
	${\tt PAmmonia}$	RKjeldah:	l PKjeld	ahl	RTotal.P	PTotal.P	RCopper	PCopper
1	NA			NA		NA		NA
2	NA			NA		NA		NA
3	NA			NA		NA		NA
4	NA			NA		NA		NA
5	NA			NA		NA		NA
6	NA			NA		NA		NA
RIron PIron Calcium Chloride								
1		NA S	95	280				
2		NA I	NΑ	NA				
3		NA 4	12	106				
4		NA 12	21	435				
5		NA 15	50	512				
6		NA 13	38	510				

2 Summarize the Sample Data

In general, it is desirable, but not necessary, to subset the data before proceeding with the analysis of a subset of the constituents. Before these data are subsetted, the FLOW column must be created. The flow data are in two columns QI, the flow at the time of the sample; and QD, the mean flow on the day of the sample. The coalesce function in the smwrBase package can used to select the non-missing value for flow.

For censored data, which includes left- and multiply-cenosred data, the response variable must be converted to class "qw." The use of this class facilitates censored data analysis. The convert2qw function in the smwrQW package can be used to convert these data. The conversion requires at least 2 columns, one for the value and one for the remark code. For these data, columns beginning with "P" contain the value and columns beginning with "R" contain the remark code; the matching suffixes define the pair. This naming scheme is known as the Booker convention. Note that USGS data retrieved using the importQW function have much more meta information and do not need conversion.

```
> # Compute FLOW, the coalese function is in smwrBase
> EstrendSub <- transform(EstrendSub, FLOW=coalesce(QI, QD))
> # Convert, the default scheme is "booker"
> # The convert2qw function is in smwrQW
> EstrendSub.qw <- convert2qw(EstrendSub)
> # Create the subset, the Pcolumn name is preserved
> Metals <- subset(EstrendSub.qw, select=c("STAID", "DATES", "FLOW",
                                            "PIron", "PCopper"))
> # Rename metals to remove the leading P
> names(Metals)[4:5] <- c("Iron", "Copper")
> # Show the first few rows of the data
> head(Metals)
     STAID
                DATES FLOW Iron Copper
1 07227500 1968-10-01
                        7.6
                              NA
                                     NA
2 07227500 1968-10-03
                        5.3
                              NA
                                     NA
3 07227500 1968-10-16 532.0
                                     NA
                              NΑ
4 07227500 1968-10-19 17.0
                                     NA
5 07227500 1968-11-01 17.0
                                     NA
                              NΑ
6 07227500 1968-12-01
                              NΑ
```

The sampReport function creates a simple PDF file that contains a report of the sample date ranges and graph of samples for each site. It can be used to help define the starting and ending date ranges for the trend tests as well as identifying sample gaps and other sampling issues. However, sampReport only reports the dates in the data set, it does not know about any missing samples. To get an accurate count of the samples, missing values across all metal columns need to be removed. The na.omit function cannot be used because it would remove rows where there were any missing values.

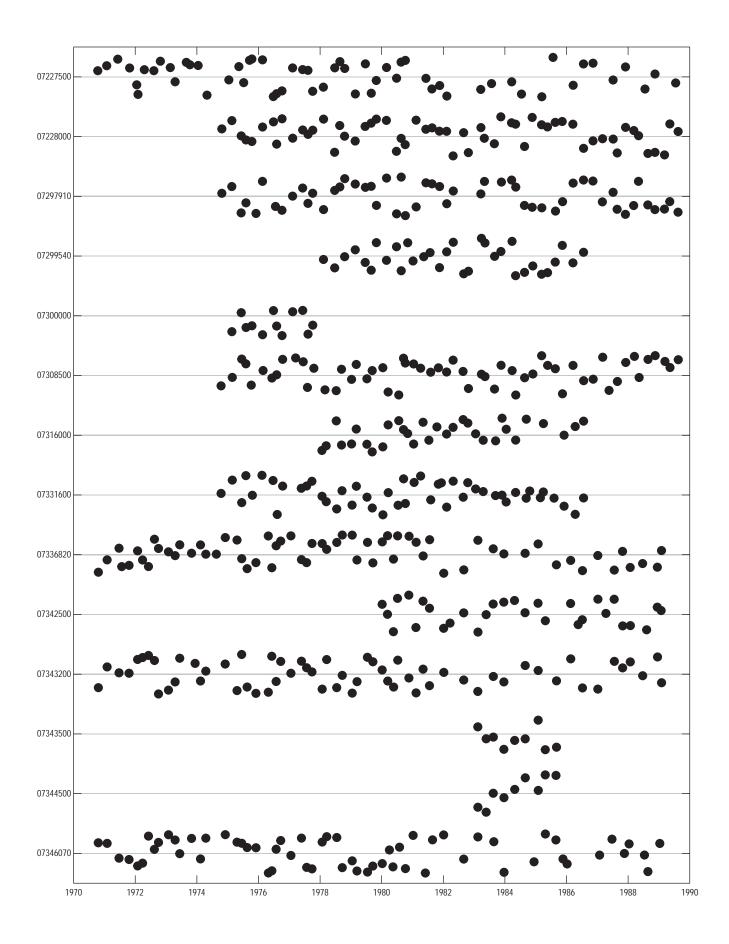
```
> # Subset the data and show first few lines
> Metals <- subset(Metals, !(is.na(Iron) & is.na(Copper)))
> head(Metals)
```

```
STAID
                 DATES FLOW Iron Copper
116 07227500 1970-10-14
                         7.4
                               38
                                       4
123 07227500 1971-01-28 38.0
                               34
                                       6
137 07227500 1971-06-08 31.0
                               30
                                       4
156 07227500 1971-10-27 104.0
                               40
                                       5
167 07227500 1972-01-19 84.0
                               20
                                       4
169 07227500 1972-02-03 30.0 720
                                      25
```

- > # Create the report
- > sampReport(Metals, DATES="DATES", STAID="STAID", file="MetalsSampling")

The call to sampReport returns the file name invisibly (MetalsSampling.pdf). Because it is a full-size portrait PDF file, it is inserted here with compressed pages. The report gives the actual begin and end dates for sampling and the graph shows the sampling dates for each station. It is easy to see that only 14 stations were sampled for metals within the analysis period and the actual sampling at each station varied widely.

	STAID	FirstSamp	LastSamp	NumSamp
1	07227500	1970-10-14	1989-07-18	58
2	07228000	1974-10-24	1989-08-16	59
3	07297910	1974-10-24	1989-08-15	54
4	07299540	1978-02-10	1986-07-22	33
5	07300000	1975-02-20	1977-10-06	12
6	07308500	1974-10-15	1989-08-18	59
7	07316000	1978-01-23	1986-07-22	32
8	07331600	1974-10-15	1986-07-22	48
9	07336820	1970-10-21	1989-01-31	63
10	07342500	1980-01-07	1989-01-27	30
11	07343200	1970-10-21	1989-02-01	61
12	07343500	1983-02-14	1985-09-04	9
13	07344500	1983-02-15	1985-08-29	9
14	07346070	1970-10-20	1989-01-11	61



3 Set up the Project

The user must balance the need to include as many stations as possible and the targeted time frame for the trend estimation. For these data, 4 stations have complete record beginning in October, 1970, but 3 additional stations have complete records beginning in October, 1974. This example will use the analysis period beginning in October, 1974 and ending in September, 1989.

The setProj function sets up the trend estimation project. There are many arguments to setProj, see the documentation for details. The constituent names or response variable names are referred to as Snames in keeping with the names used in the original ESTREND.

After projects have been set up, the user can get a list of the projects by using lsProj or can specify a project to use with useProj. The function useProj must be used to continue working on a project after the user quits from the R session.

[1] "metals"

The setProj function creates a folder in the users workspace with that name. That folder contains R data that are updated after each successful call to an analysis function in restrend. Table 1 describes the data created in this example's call to setProj. Any object of class "matrix" or "by" are indexed by station and sname.

Table 1. The data created by setProj.

Name	Class	Description				
estrend.cl list		A record of the calls to analysis functions.				
estrend.cn	matrix	A description of the censoring. May be "none," "left," or "multiple."				
estrend.cp	matrix	The percent of observations that are left-censored.				
estrend.df	by	The dataset, contains STAID, DATES, FLOW, and				
		the response variable.				
estrend.in	list	Information about the project, such as the start and				
		end dates and the names of columns in each dataset.				
estrend.st	matrix	The status for each station and sname. Must be				
		"OK" to continue with the trend analysis.				

It is useful to verify which stations and snames will be analyzed. The user need only enter the name of the R data object in the console. The stations listed as "OK" matches what we expect from the sample report.

```
07299540 "short record" "short record"
07300000 "too few data" "too few data"
07308500 "OK" "OK"
07316000 "short record" "short record"
07331600 "short record" "short record"
07336820 "OK" "OK"
07342500 "short record" "short record"
07343200 "OK" "OK"
07343500 "too few data" "too few data"
07344500 "too few data" "too few data"
07346070 "OK" "OK"
```

4 Tobit Trend Test

These data are now ready for the Tobit trend test. The function tobitTrends executes the trend test on all valid combinations of stations and snames. It can also execute the test on subsets if some changes need to be made. By default, the data are log-transformed and flow (also log-transformed) and first-order Fourier terms for seasonality are included in the regression analysis. The variable the describe the annual trend is called .Dectime and is always the last variable in the report.

The tobitTrends function also creates a PDF file that contains the result of the analysis and diagnostic graphs on each page. Most trends are very small for these data; only the report for Iron at 07346070 is shown. That is the only trend significant at the 0.05 level. The partial residual plot of trend shows some nonlinearity, but the reported slope is a good estimate of the average trned over the analysis period.

```
> # Trend tests, accepting default
> tobitTrends()
```

[1] "metals_tb.pdf"

```
Call:
censReg(formula = log(Iron) ~ log(FLOW) + fourier(.Dectime, 1) +
     .Dectime)
Coefficients:
                                     Estimate Std. Error z-score
(Intercept)
                                   -179.27196
                                                  40.80412 -4.3935
                                                    0.04539 1.9824
log(FLOW)
                                      0.08999
fourier(.Dectime, 1)sin(k=1)
                                     -0.12182
                                                    0.18443 -0.6606
                                                    0.12535 -3.4458
fourier(.Dectime, 1)cos(k=1)
                                     -0.43195
.Dectime
                                                    0.02058 4.5328
                                      0.09330
                                   p-value
(Intercept)
                                    0.0000
log(FLOW)
                                    0.0400
fourier(.Dectime, 1)sin(k=1)
fourier(.Dectime, 1)cos(k=1)
                                   0.0006
Estimated residual standard error (Unbiased) = 0.56
Distribution: normal
Number of observations = 45, number censored = 0 (0 percent)
Loglik(model) = -35.11 Loglik(intercept only) = -46.63
  Chi^{1.5}square = 23.05, degrees of freedom = 4, p-value = 0.0001
Computation method: AMLE
     1.0
     0.5
                                                                 Standardized Residuals
 Working Residuals
     -0.5
     -1.0
       5.0
                                       6.5
                                                  7.0
                                                             7.5
                                 Fitted
                                                                                            Normal Quantiles
         Second order polynomial test for linearity: p=0.0072
                                                                        Second order polynomial test for linearity: p=0.7426
     186.5
                                                                     1.5
     186.0
                                                                     1.0
     185.5
                                                                     0.5
 Partial Residual
                                                                 Partial Residual
    185.0
                                                                     0.0
     184.5
                                                                     -0.5
     184.0
     183.5
                                                                     -1.0
                                        1,984
                    1,978
                                 1,982
                                                     1,988
                                                                                -2
                                                                                         0
       1,974
             1,976
                          1,980
                                               1,986
                                                            1,990
                                                                                                                              8
                                                                                                                     6
                                                                                               log(FLOW)
                                 .Dectime
```

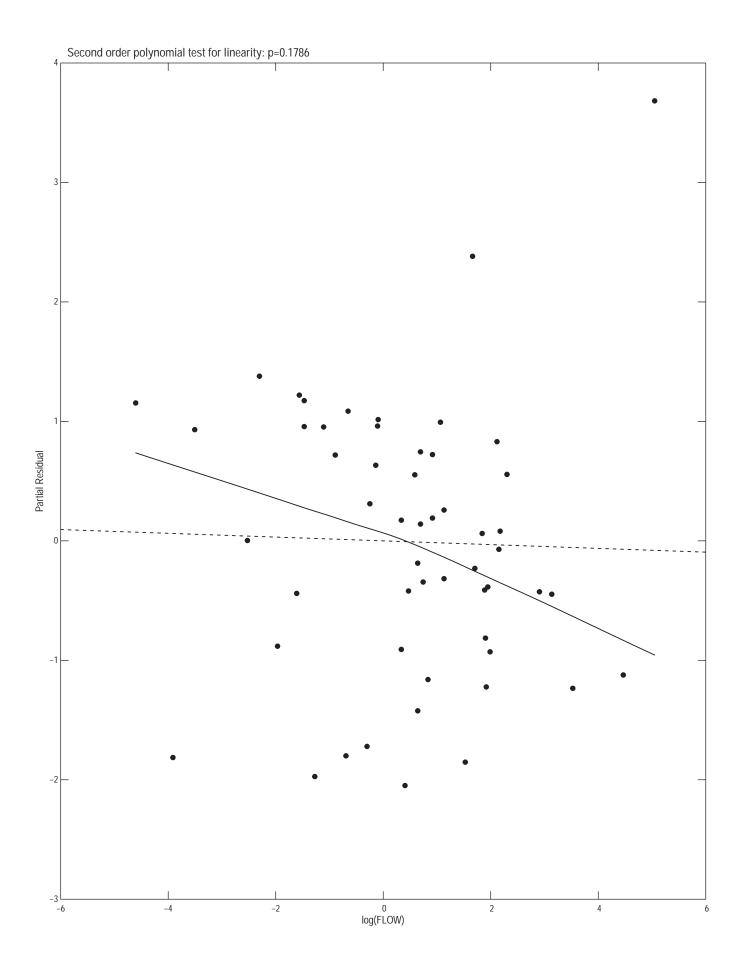
07346070 Iron

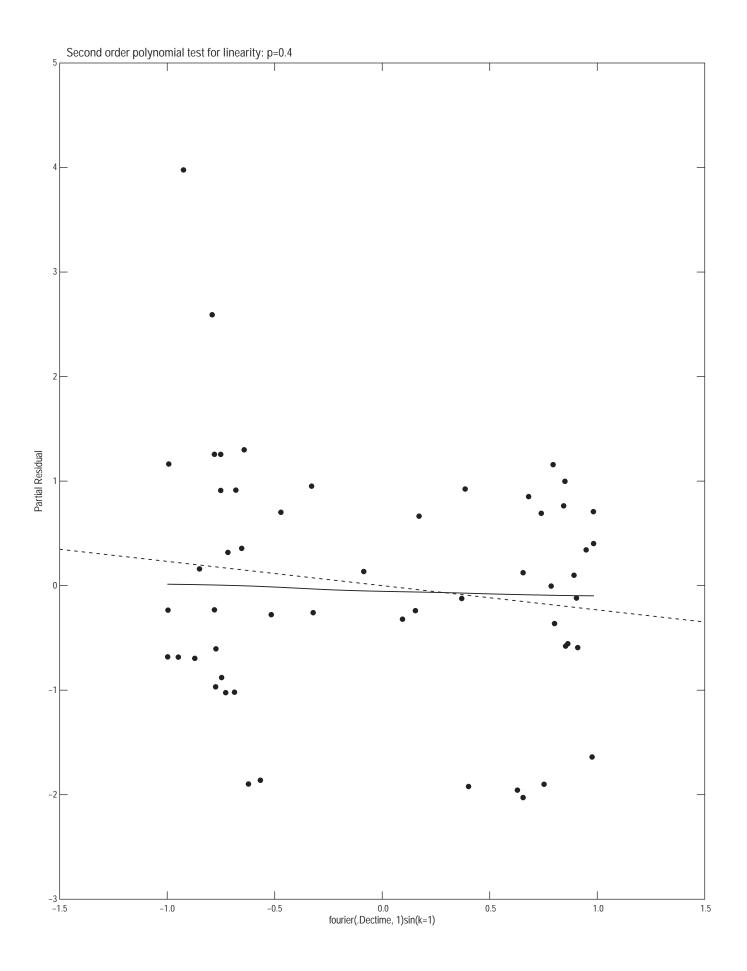
The diagnostic plots should be reviewed for verify the basic assumptions of linear regression—linearity of fit, uniformity of residuals, and normality of residuals. Note that the linearity and uniformity of residuals can be deceptive in the residuals vs. fit graph because of discrete values and censoring.

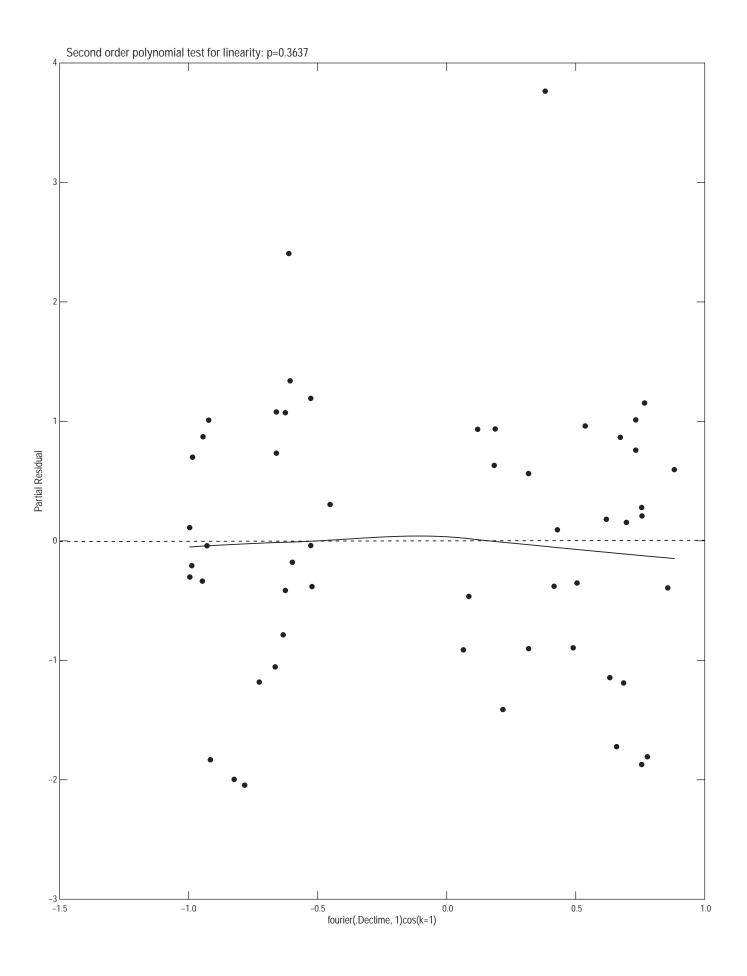
A more complete suite of diagnostic plots can be obtained using the plotTT function. For this example, there appears to be a highly influential observation in the analysis for Iron at station 07297910. The diagnostic plots reveal nothing unusual, except that it occurs very early in the record and at the largest flow.

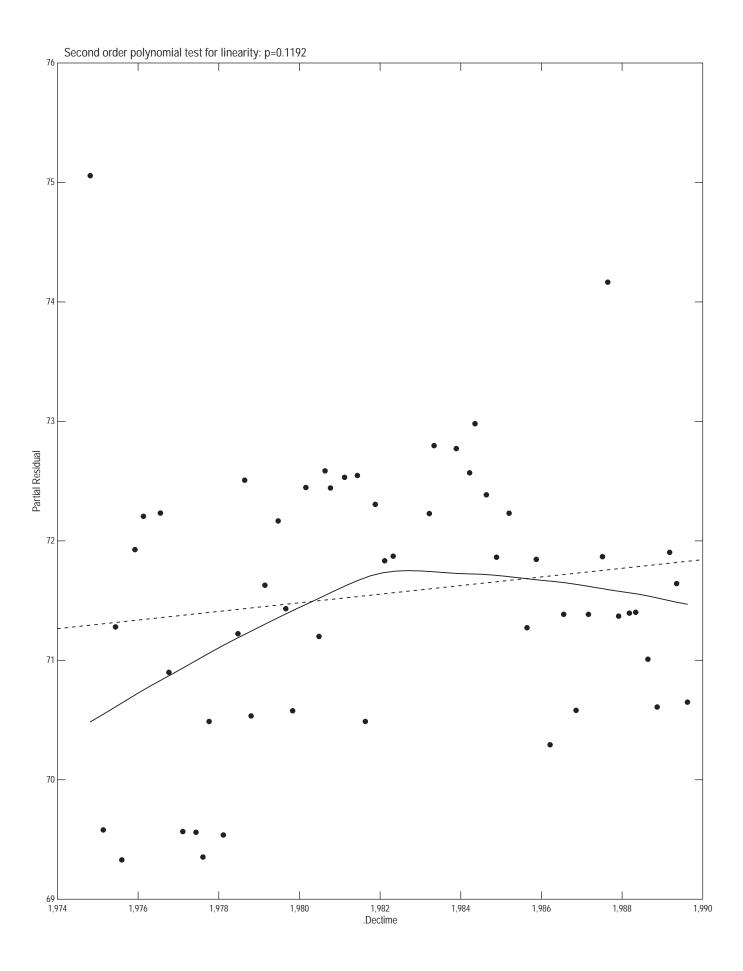
```
> # Trend tests, accepting default
> plotTT(Station="07297910", Sname="Iron", device="pdf")
```

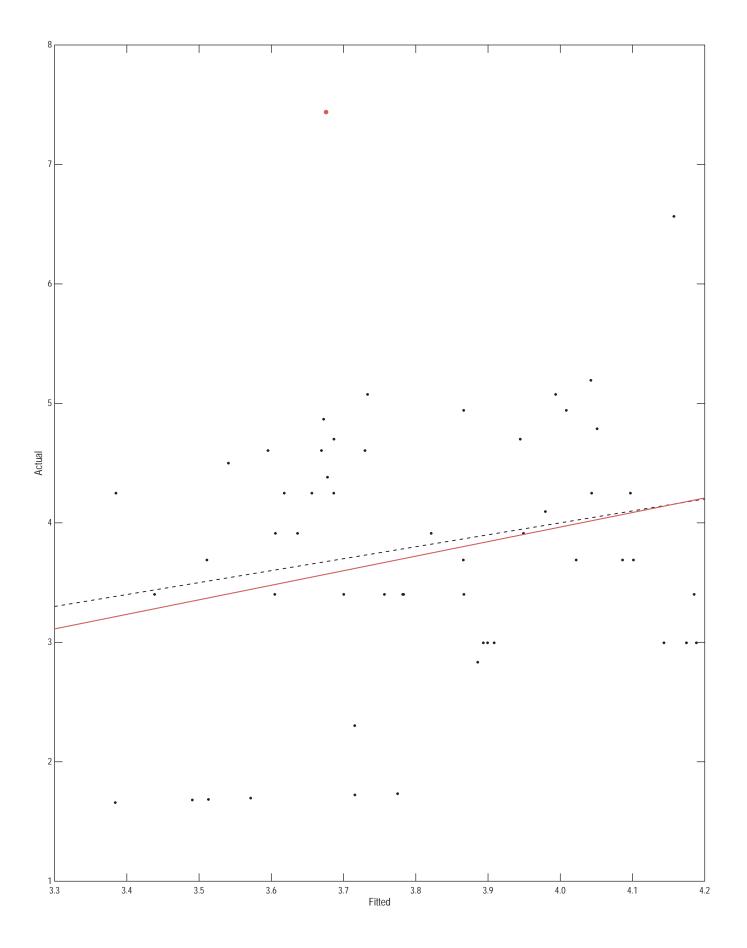
[1] "X07297910_Iron.pdf"

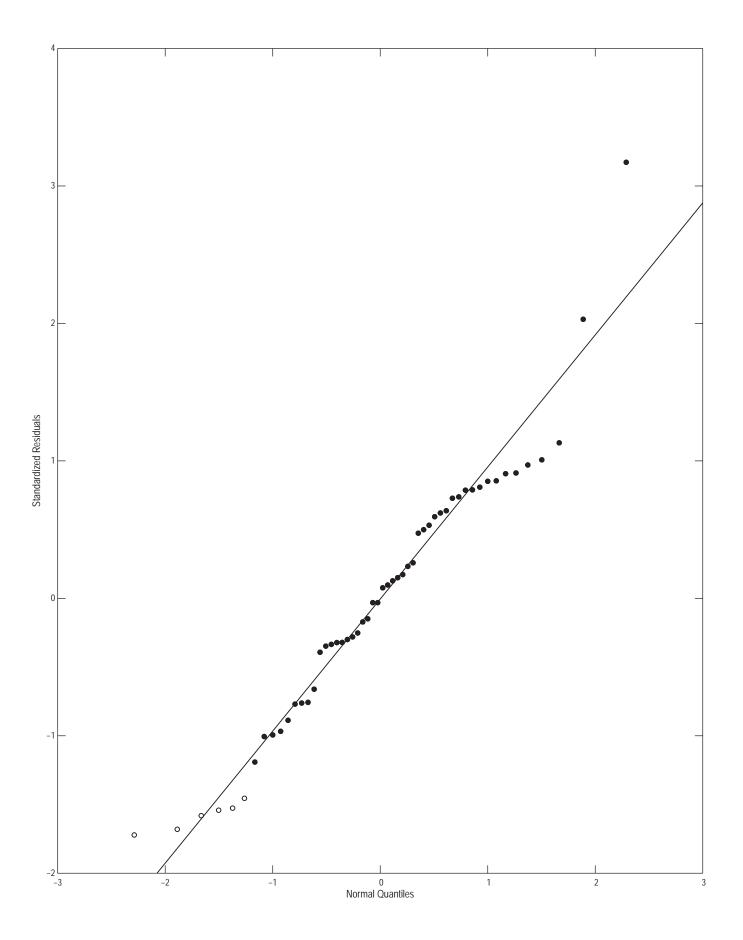


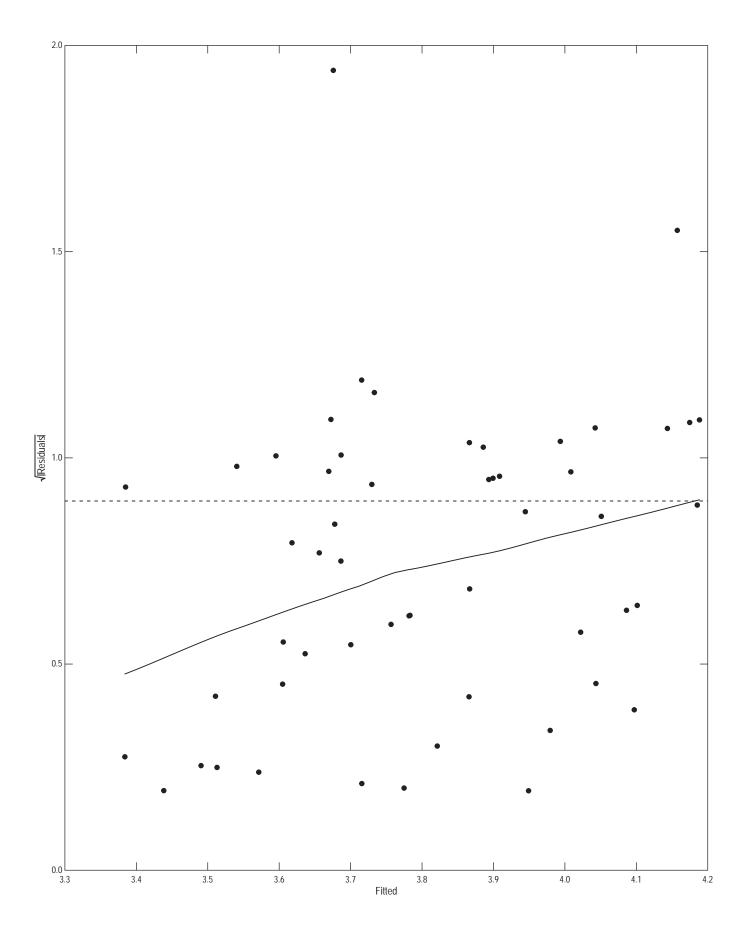


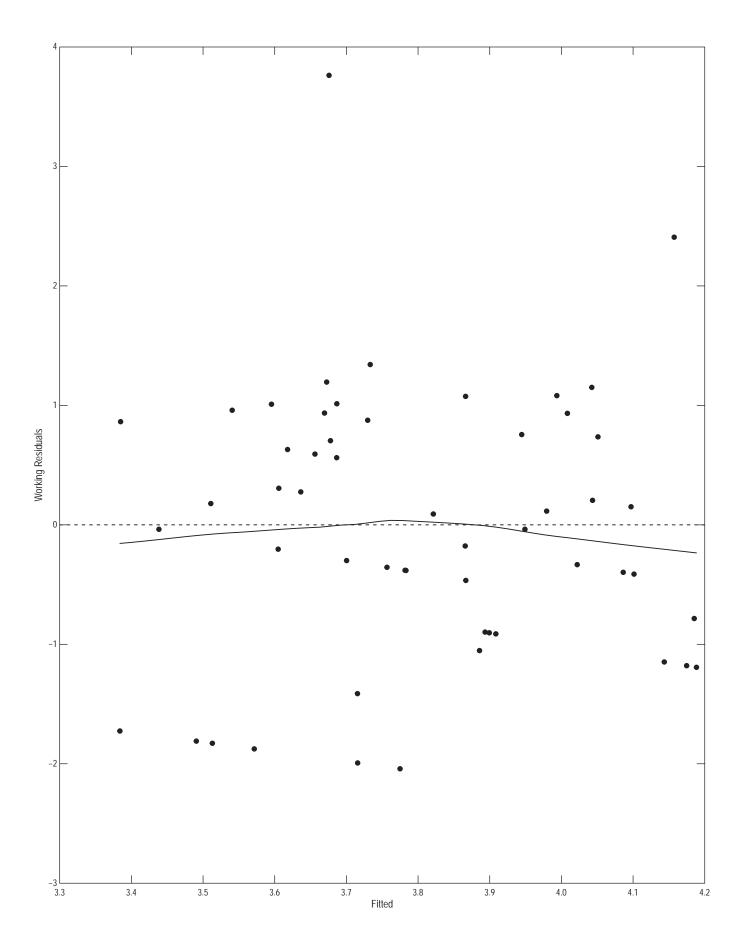


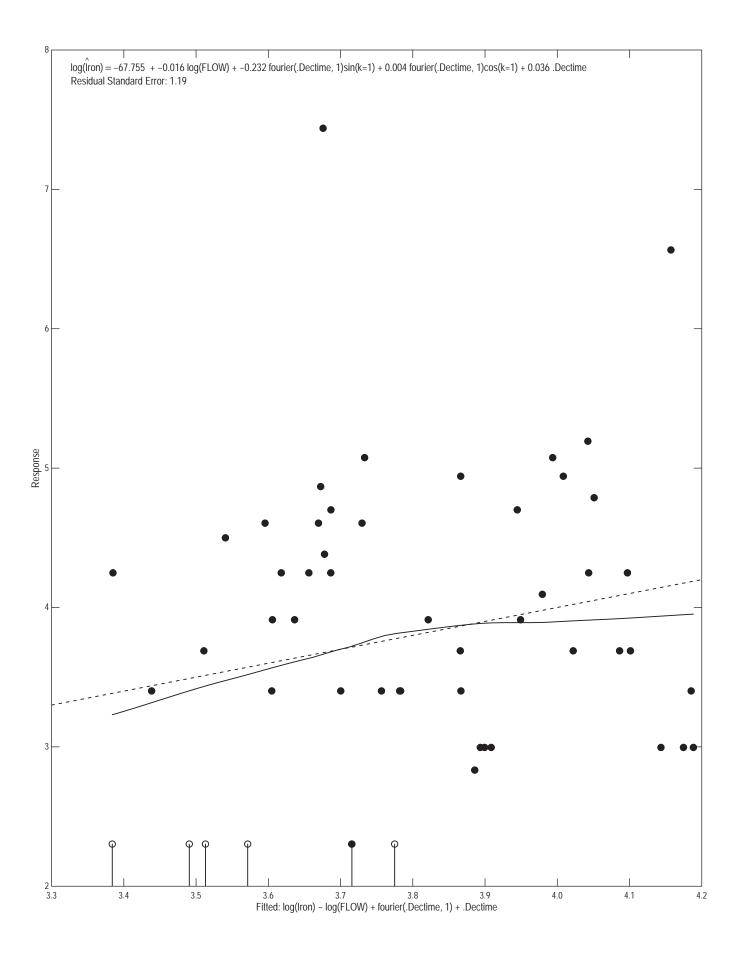












5 Trend Results

When completed, or to check on intermediate results, the estimated trends can be extracted using the getTrends function. By default, all stations and snames are extracted. The output dataset is explained in the documentation for getTrends. The user has the option to set a significance level to determine whether there is a significant trend, the default level is 0.05.

```
> # get the trends
> metals.tnd <- getTrends()
> print(metals.tnd)
```

	a	ъ		N 37	N G		D 11 3
		Response					-
1	07227500			14.50513	NA	42	20
2	07227500			14.50513		42	2
3	07228000			14.81177	NA	58	20
4	07228000			14.81177	NA	58	1
5	07297910			14.80903	NA	54	40
6	07297910			14.80903	NA	54	1
7	07308500	Iron	Tobit	14.84189	NA	55	30
8	07308500	Copper	Tobit	14.84189	NA	55	2
9	07336820	Iron	Tobit	14.16016	NA	46	30
10	07336820	Copper	Tobit	14.16016	NA	46	1
11	07343200	Iron	${\tt Tobit}$	14.16564	NA	46	30
12	07343200	Copper	${\tt Tobit}$	14.16564	NA	46	2
13	07346070	Iron	${\tt Tobit}$	14.10541	NA	45	400
14	07346070	Copper	${\tt Tobit}$	14.10541	NA	46	2
	T	rend Tre	end.pct	P. v	value Tr	end.di	r.
1	-0.540824	4909 -2.70	0412454	1 5.357736	6e-01	non	ıe
2	0.001973	3088 0.09	9865438	9.953258	3e-01	non	ıe
3	0.563684	4368 2.8	1842184	1 3.58089	1e-01	non	ıe
4 0.034454725 3.44547250 1.344894e-01 none							
5	1.47045	1671 3.6	7612918	3.281969	9e-01	non	ıe
6	0.01104	1926 1.10	0419259	7.361143	3e-01	non	ıe
7	-0.019554	4571 -0.06	3518190	9.846199	9e-01	non	ıe
8	0.068046	6237 3.40	0231183	3 2.899362	2e-01	non	ıe
9	-0.518956	6445 -1.72	2985482	2 5.292452	2e-01	non	ıe
10		6056 4.48				non	ıe
11	0.936588	3462 3.13	2196154	4.079293	3e-01	non	ıe
12		4124 2.2				non	ıe
	39.11604						ıp
14		3859 5.7					ıp
	3.22200				-	·	T.

6 Further Remarks

Because trend analysis is not necessarily a straightforward process, but requires user assessments at several points in the process, it is not necessarily a good idea to simply create scripts and run them without any user review and interaction. To overcome recording the steps in a script, the functions in restrend record all changes to the projects database in a list called estrend.cl. It can be viewed at any time simply by entering estrend.cl in the console window. It can be saved with the data to ensure that the trend analysis is reproducible.

References

- [1] Lorenz, D.L., in preparation, restrend: an R package for EStimate TRENDs: U.S. Geological Survey Open File Report, ? p.
- [2] Schertz, T.L., Alexander, R.B., and Ohe, D.J., 1991, The computer program Estimate TREND (ESTREND), a system for the detection of trends in water-quality data: U.S. Geological Survey Water Resources Investigations Report 91-4040, 72 p.