

# rockchalk package

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# 2013



# Outline

- 1 Introduction
- 2 Data
- 3 Outreg
- 4 Plots
  - Categorical modx
  - Numeric moderator
- 5 Free Lunch
- 6 Conclusions
- 7 Guessing

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# Thanks for Joining

Thanks to Ray DiGiacomo, Jr & OC RUG for organizing

Downloads:

<http://pj.freefaculty.org/guides> {all my lectures on anything}

[.../Rcourse/rockchalk-2013](http://pj.freefaculty.org/Rcourse/rockchalk-2013) {this lecture, source code, LyX doc, etc}

<http://pj.freefaculty.org/R>: Rtips, links to other R stuff

# Why Make a Package?

- Avoid a riot after an influx of 40 MA-bound behavioral scientists into my regression class
  - Honestly, I'd rather teach R programming, but
  - I can understand the view that statistics exists apart from R
- Package has “convenience” functions for
  - Me preparing lectures
  - Them doing papers (with nice looking graphs!)
- I had distributed functions before, but never made a package



# What do you expect in rockchalk?

- Functions for difficult/tedious/hard-to-teach chores
- Verbose documentation, (too) many examples
- vignettes
  - “*rockchalk*”: Discussion & demonstration of package
  - “*Rchaeology*”: Deep insights into R programming I accumulate while working on the package
  - “*Rstyle*”: The style manual I wish R Core would adopt
- Hidden value added: the examples folder in the package install directory includes some special educational R examples (look for `noWords-001.R` and `centeredRegression.R`)



## Where is the hard work in version 1.8?

- `predictOMatic()`. *Flexible* way to demonstrate marginal effects of predictors. Goal: make it easy to understand regression as a translation of inputs into predicted values (and uncertainty)
- Scan fitted regressions, create newdata objects with possible predictor values (“divider” algorithms to create focal values for consideration).

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# Make a Presentable Table Describing The Data

- Assignment: create a summary table for your research article
- R's `summary()`
  - does not include diversity estimates
  - does not separate numeric from factor variables in the report
  - does not provide output in a usable format
- `rockchalk summarize()`
  - does

## Example

```
(datsum <- summary(dat))
```

income	educ	age	religion	gender
Min. : -56816	Min. : 2.00	Min. : 9.00	cath : 177	female : 532
1st Qu.: -2225	1st Qu.: 8.00	1st Qu.: 19.00	jewish : 87	male : 468
Median : 10565	Median : 10.00	Median : 22.00	muslem : 94	
Mean : 10473	Mean : 10.02	Mean : 22.04	other : 294	
3rd Qu.: 23772	3rd Qu.: 12.00	3rd Qu.: 25.00	prot : 169	
Max. : 77189	Max. : 21.00	Max. : 37.00	roman : 105	
NA's : 80	NA's : 40		NA's : 74	

- Can you wrestle that into a paper? I can't! It has text and values combined

```
datsum[,1]
```

```
"Min. : -56816 " "1st Qu.: -2225 " "Median : 10565 " "Mean : 10473 " "3rd
Qu.: 23772 " "Max. : 77189 " "NA's : 80 "
```

- Default output from summarize separates numerics & factors, alphabetizes

## Example ...

```
datsum2 <- summarize(dat)
```

The result object `datsum2` is a list with 2 parts, a numeric matrix part and a factor variable display.

- The numerics are a matrix, easy to take rows or columns to put into a paper

```
datsum2$numerics
```

	age	educ	income
0%	9.000	2.000	-56820
25%	19.000	8.000	-2225
50%	22.000	10.000	10570
75%	25.000	12.000	23770
100%	37.000	21.000	77190
mean	22.040	10.020	10470
sd	4.556	3.056	19630
var	20.760	9.337	385400000
NA's	0.000	40.000	80
N	1000.000	1000.000	1000



## Example ...

- The factors are a separate list

```
datsum2$ factors
```

gender		religion	
female	: 532.000	other	: 294.0000
male	: 468.000	cath	: 177.0000
NA's	: 0.000	prot	: 169.0000
entropy	: 0.997	roman	: 105.0000
normedEntropy:	0.997	(All Others)	: 181.0000
N	:1000.000	NA's	: 74.0000
		entropy	: 2.4414
		normedEntropy:	0.9445
		N	:1000.0000

- Indicators of central tendency and dispersion are included in both displays
- Try `summarizeNumerics()` and `summarizeFactors()` to get just one or the other.



## Sidenote: recoding a factor

- Note the religion variable has levels “cath” and “roman”, which was a data entry error. Catholic and Roman Catholic represent the same idea
- Did you ever try to write R code to fix that (without killing yourself)?
- Try `rockchalk::combineLevels()`:

```
dat$religion2 <- combineLevels(dat$religion ,
  c("cath", "roman"), "cath")
```

The original levels cath jewish muslem other  
prot roman  
have been replaced by jewish muslem other prot  
cath



## Sidenote: recoding a factor ...

```
table(dat$religion2 , dat$religion , dnn = c("religion2", "
religion"))
```

	religion					
religion2	cath	jewish	muslem	other	prot	roman
jewish	0	87	0	0	0	0
muslem	0	0	94	0	0	0
other	0	0	0	294	0	0
prot	0	0	0	0	169	0
cath	177	0	0	0	0	105

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## Need a Nice Looking Regression Table?

- Each student should not invent a unique report format for regressions.
- MS Word users especially tempted to “finger paint” with fonts and formats.
- Solution: provide usable  $\text{\LaTeX}$  tables (added benefit: bait to get them to use  $\text{\LaTeX}$ )
- rockchalk-1.8 provides HTML backend as well (compromise with reality)



## For many years, outreg was a function in search of a package

- Dave Armstrong (then at U. Maryland student) gave me the outreg idea 10 years ago
- I wrote up a function that more-or-less worked, distributed it, revised it as my R programming skills improved
- I didn't know there was "outreg" module for Stata. . . .



## outreg example usage

I fit a regression using a subset of the American National Election Study 2004 (ICPSR), which I called “mydta1”

```
mod1age <- lm(th.bush.kerry ~ V043250, data =  
  mydta1)  
outreg(mod1age, tight = FALSE, modelLabels =  
  c("Age as Predictor"))
```



## Produces this LaTeX Markup

```

\begin{tabular}{*{3}{l}}
\hline
& \multicolumn{2}{c}{Age as Predictor} & \\
& Estimate & (S.E.) & \\
\hline
\hline
(Intercept) & -6.841 & (4.596) & \\
V043250 & 0.184* & (0.092) & \\
\hline
N & 1191 & & \\
RMSE & 53.885 & & \\
$R^2$ & 0.003 & & \\
\hline
\hline
\multicolumn{2}{l}{{*} p \le 0.05} & & \\
\end{tabular}

```



## Which LaTeX Renders as

	Age as Predictor	
	Estimate	(S.E.)
(Intercept)	-6.841	(4.596)
V043250	0.184*	(0.092)
N	1191	
RMSE	53.885	
$R^2$	0.003	

\* $p \leq 0.05$

My terminology:

tight = FALSE  $\Rightarrow \hat{\beta}$  and  $std.err(\hat{\beta})$  are side by side

tight = TRUE  $\Rightarrow \hat{\beta}$  and  $std.err(\hat{\beta})$  are vertically aligned.



## Add Gender

```
## Run a new regression  
mod2age <- lm(th.bush.kerry ~ V043250 +  
  V041109A, data = mydata1)  
## Put 2 regressions in same table  
outreg(list(mod1age, mod2age), tight = TRUE,  
  modelLabels = c("Age Only", "Age With  
  Gender"))
```

NB: tight = TRUE





## Output To LaTeX

	Age Only Estimate (S.E.)	Age With Gender Estimate (S.E.)
(Intercept)	-6.841 (4.596)	4.628 (6.527)
V043250	0.184* (0.092)	0.191* (0.092)
V041109A	.	-7.713* (3.123)
N	1191	1191
RMSE	53.885	53.77
$R^2$	0.003	0.008
adj $R^2$	0.003	0.007

\* $p \leq 0.05$



## Alternative way to specify model labels (rockchalk 1.8)

```
outreg(list("Age Only" = mod1age, "Age With  
Gender" = mod2age), tight = FALSE)
```

Perhaps more coherent usage: keep labels with models in a list

# Output To LaTeX

	Age Only		Age With Gender	
	Estimate	(S.E.)	Estimate	(S.E.)
(Intercept)	-6.841	(4.596)	4.628	(6.527)
V043250	0.184*	(0.092)	0.191*	(0.092)
V041109A	.		-7.713*	(3.123)
N	1191		1191	
RMSE	53.885		53.77	
$R^2$	0.003		0.008	
adj $R^2$	0.003		0.007	

\* $p \leq 0.05$





## Beautify Variable Labels

```
outreg(list("Age Only" = mod1age, "Age With  
Gender" = mod2age), tight = TRUE,  
varLabels = list("V043250" = "Age", "  
V041109A" = "Gender"))
```

Quotation marks optional before equal sign in list; this works too

```
outreg(list("Age Only" = mod1age, "Age With  
Gender" = mod2age), tight = FALSE,  
varLabels = list(V043250 = "Age",  
V041109A = "Gender"))
```

Not necessary to provide new labels for all variables





## My Beautiful Table with Lovely Variable Labels

	Age Only		Age With Gender	
	Estimate	(S.E.)	Estimate	(S.E.)
(Intercept)	-6.841	(4.596)	4.628	(6.527)
Age	0.184*	(0.092)	0.191*	(0.092)
Gender	.		-7.713*	(3.123)
N	1191		1191	
RMSE	53.885		53.77	
$R^2$	0.003		0.008	
adj $R^2$	0.003		0.007	

\* $p \leq 0.05$



## Quick R style comment: My opinion

- Students often have urge to rename variables in the analysis itself, to create new `dat$gender` and `dat$age`
- I urge them to resist the temptation
- In a team setting, everybody has same input variables with names like `V234234`, cooperation is frustrated when everybody renames everything
- However, in output, no reader wants to see `V234234`

# What is this Good For?

- Good-enough tables in lectures & term papers
- Possible to “script” together a lot of separate estimates for a lot of different datasets
- Especially when the students start to think they know everything, show *I'm still smarter than you*:
  - <http://pj.freefaculty.org/R/gloating/test2>
  - <http://pj.freefaculty.org/guides/stat/Regression/Multicollinearity/Multicollinearity-1-lecture.pdf>



## Recent updates to outreg

- I get more emails about `outreg()` than any of the other functions. People want more and more features.
- Compromises so far allow customization of:
  - model “header” labels and variable names
  - the selection of “goodness of fit” indicators in the bottom of the table
  - choice of alpha levels (Previously, I first refused p-values, then insisted only 0.05).
  - HTML output (next slide)

## outreg can create html file output

- This is a brand new feature in outreg 1.8 (June, 2013)
  - outreg2HTML() receives outreg results and converts into Web markup.
  - Wrestle that into Word however you like.
    - open the html document File -> Open
    - view the html document in a web browser, copy & paste manually into word (use paste Special HTML).
  - Not as nice looking or as automatic as  $\text{\LaTeX}$ , but I may try harder in future

## HTML output

```
or1 <- outreg(list(mod1age, mod2age), tight =  
  TRUE, modelLabels = c("Age Only", "Age  
  With Gender"))  
outreg2HTML(or1, file = "or1-test.html")
```

That creates a file, "or1-test.html". See if your web browser can open it. See if Word can open that. I uploaded a copy you can inspect: <http://pj.freefaculty.org/R/or1-test.html>

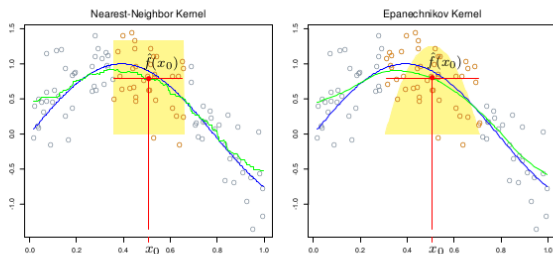
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# I want it to be easy to make scatterplots with Predicted Values

## 192 6. Kernel Smoothing Methods



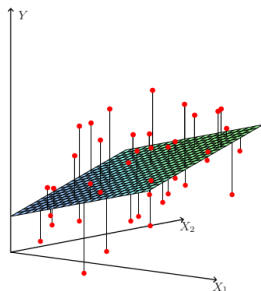
**FIGURE 6.1.** In each panel 100 pairs  $x_i, y_i$  are generated at random from the blue curve with Gaussian errors:  $Y = \sin(4X) + \varepsilon$ ,  $X \sim U[0, 1]$ ,  $\varepsilon \sim N(0, 1/3)$ . In the left panel the green curve is the result of a 30-nearest-neighbor running-mean

T. Hastie, R. Tibshirani, J. Friedman, *Elements of Statistical Learning: Data Mining, Inference, And Prediction, 2ed* (Springer,



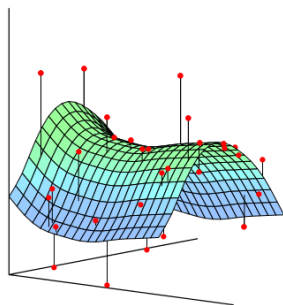
# Especially in 3D

3.2 Linear Regression Models and Least Squares 45



**FIGURE 3.1.** Linear least squares fitting with  $X \in \mathbb{R}^2$ . We seek the linear function of  $X$  that minimizes the sum of squared residuals from  $Y$ .

2.6 Statistical Models, Supervised Learning and Function Approximation

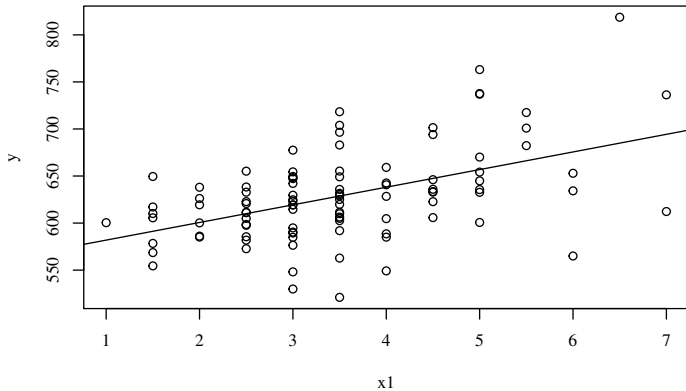


## abline is an R staple

- Everybody has done this. (Right?)

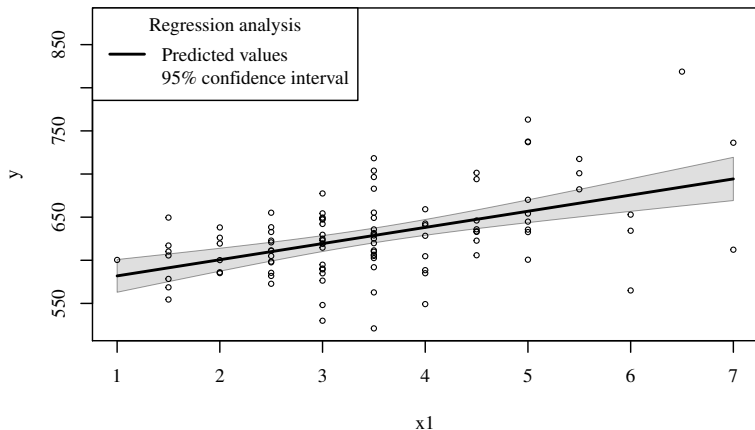
```
mod1 <- lm(y ~ x1, data = dat)
plot(y ~ x1, data = dat)
abline(mod1)
```

## abline



## I'd rather look at this plot

```
ps1 <- plotSlopes(mod1, plotx = "x1",  
  interval = "confidence")
```



## abline's fatal flaws

- Suppose the regression model is

```
mod4 <- lm(y ~ x1 + x2 + x3, data = dat)
```

```
mod2 <- lm(y ~ log(x1), data = dat)
```

```
mod2 <- lm(y ~ x1*x2, data = dat)
```

```
mod3 <- glm(y ~ x1, data = dat, family = "binomial")
```

- Common answer: abline is an epic fail.



## I have taught this "'easy' 3 step procedure" many times

**Step 1.** Create a "newdata" data frame that has values of the x's for which we want to calculate predictions.

**Step 2.** Use that newdata object (say, ndat) with the regression model's predict method, with syntax like

```
p1 <- predict(mod1, newdata = ndat)
```

or, if confidence intervals are desired,

```
p2 <- predict(mod1, newdata = ndat,  
              interval = "confidence")
```

Frustratingly, p1 and p2 are returned as different object types

**Step 3.** Wrestle those predicted values into a plot



## A sophisticated R user should learn to do that

- I've taught that (look for notes in <http://pj.freefaculty.org/R/WorkingExamples>), but it is too difficult
- I needed to create plots and calculate correlations as described in *Applied Multiple Regression*, by Cohen, Cohen, West, and Aiken, (Routledge, 2002). Students needed lots of R help, some calculations not trivial.
- `plotSlopes()` is the “simple-slope” routine ala CCWA, it was improvised in an emergency, `plotCurves()` & `plotPlane()` used same terminology for consistency.





# Syntax

- User fits `m1`, a multiple regression
- Then gives that to `plotSlopes()`, with arguments
  - `plotx`: The name of the variable on the horizontal axis
  - `modx`: The name of a “moderator” variable on which predicted values may depend.
  - `modxVals`: Values of the moderator for which “simple slopes” are desired
- Other arguments will be passed through to `plot()` and `predict()`
- See the rockchalk vignette.



## Difference between plotSlopes and plotCurves

- `plotSlopes()`: for linear models
  - Allows interactions (unlike `termplot()`)
  - Output object can be passed to rockchalk function `testSlopes()`
- `plotCurves()`: for nonlinear models (`lm()` or `glm()`).
  - Complete drop-in replacement for `plotSlopes()`
  - Nonlinear formulae in the predictors (succeeds where `termplot` fails)
  - Does not create object suitable for `testSlopes()`

## Example: moderator is an R factor

- $x_3$  is a predictor with values “left” and “right”
- If there are more predictors, they will be set to their central values (mean or mode) for calculation of predicted values

```
mod1 <- lm(y2 ~ x1*x3, data = dat)
ps1 <- plotSlopes(mod1, plotx = "x1", modx =
  "x3")
```





## The estimated regression is

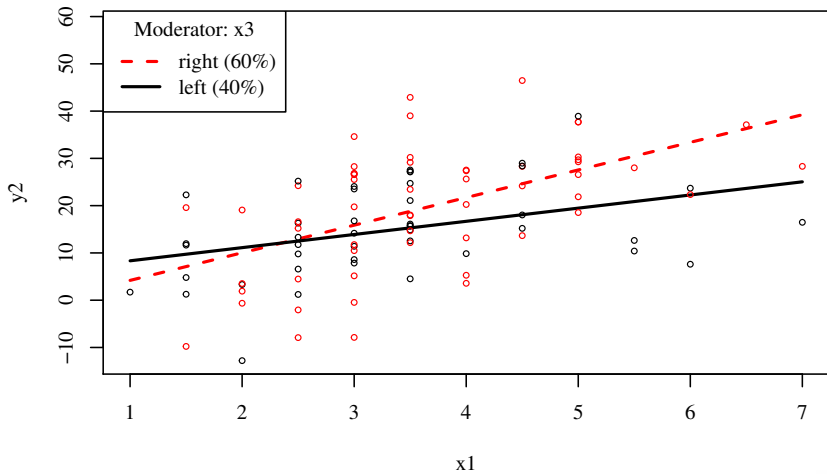
	Example Interaction	
	Estimate	(S.E.)
(Intercept)	5.549	(4.199)
x1	2.785*	(1.172)
x3right	-7.197	(6.104)
x1:x3right	3.055	(1.644)
N	100	
RMSE	10.312	
$R^2$	0.277	
adj $R^2$	0.254	

\* $p \leq 0.05$





## 2 lines, one for each value of modx

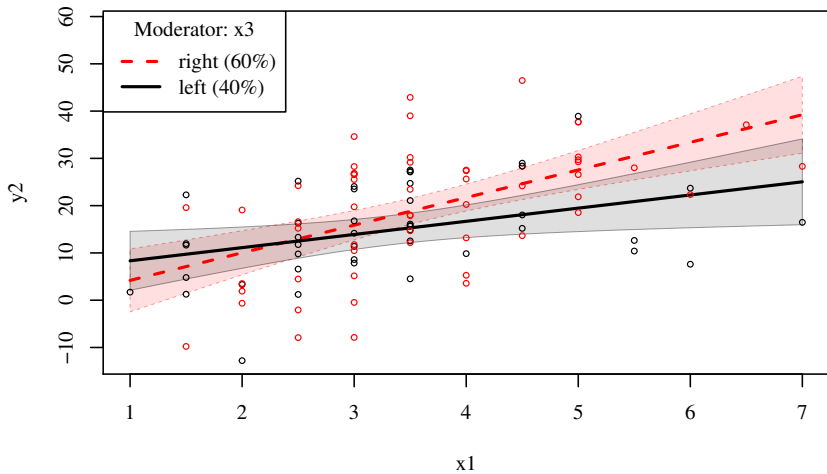


## Add confidence interval

```
ps2 <- plotSlopes(mod1, plotx = "x1", modx =  
  "x3", interval = "confidence")
```



## Add confidence interval



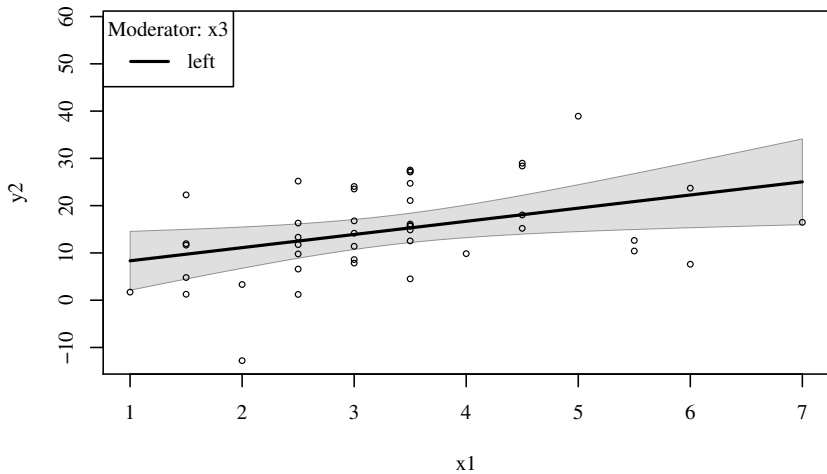
## Plot a particular group

```
ps3a <- plotSlopes(mod1, plotx = "x1", modx =  
  "x3", modxVals = c("left"), interval = "  
  confidence")
```





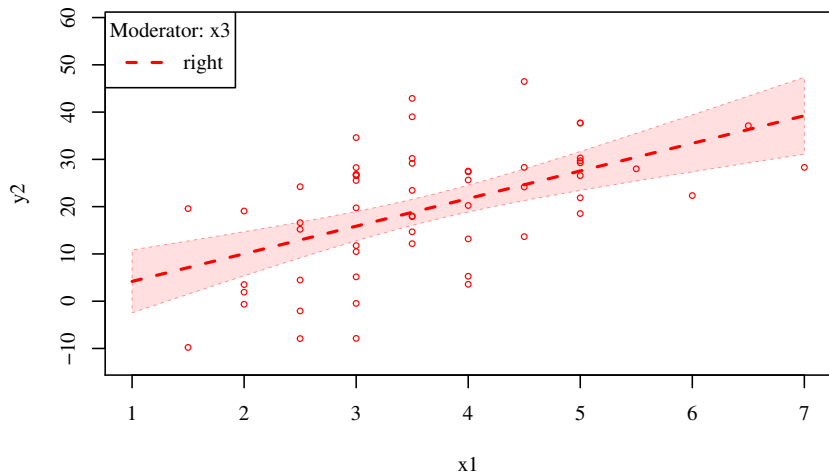
## Plot of values for "left" group



## Plot of values for "right" group

```
ps3b <- plotSlopes(mod1, plotx = "x1", modx =  
  "x3", modxVals = c("right"), interval = "  
  confidence")
```

Note my hard work to keep colors consistent





## What if the modx variable is numeric?

- When modx is numeric, then particular values need to be chosen for plotting
- Originally, I thought users would explicitly specify values, modxVals
- Have received many user requests, rockchalk 1.8 offers a variety of selection methods.

## What if the modx variable is numeric?

- psychologists generally prefer *mean – std.dev.*, *mean*, *mean + std.dev.* (or more standard deviations)
- other fields prefer quantiles, such as the 25%, 50% and 75%
- User selects either particular values or a “divider algorithm” to get this done



## Defaults

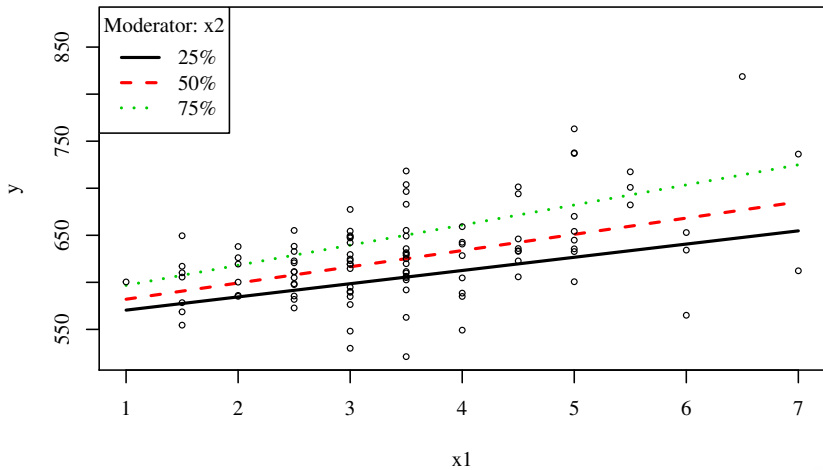
```
mod2 <- lm(y ~ x1*x2, data = dat)
ps5 <- plotSlopes(mod2, plotx = "x1", modx =
  "x2")
```

The default will select the 3 middle quartiles





## plotSlopes with numeric modx



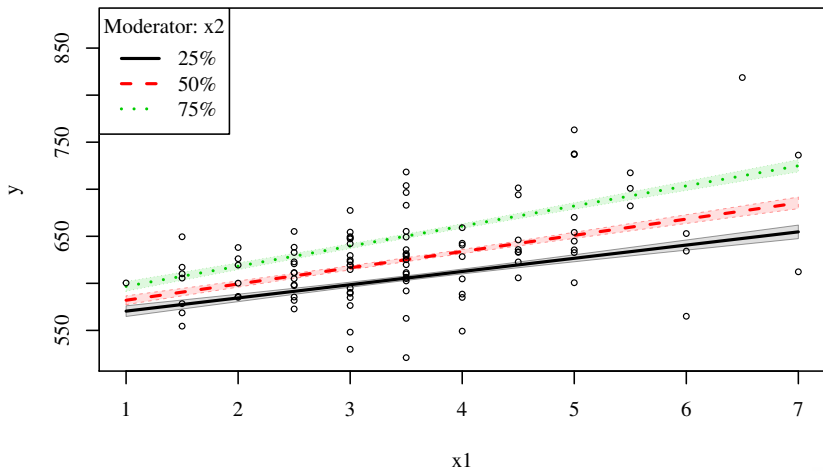
## Add confidence intervals

```
ps5 <- plotSlopes(mod2, plotx = "x1", modx =  
  "x2", interval = "confidence")
```





## plotSlopes with confidence intervals

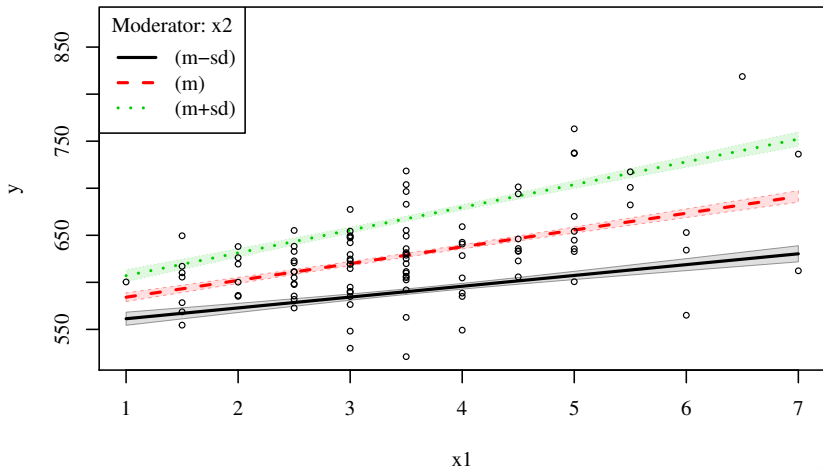


## Change the algorithm to chose modx values

```
ps7 <- plotSlopes(mod2, plotx = "x1", modx =  
  "x2", modxVals = "std.dev.", interval = "  
  confidence")
```



# std.dev, +/-

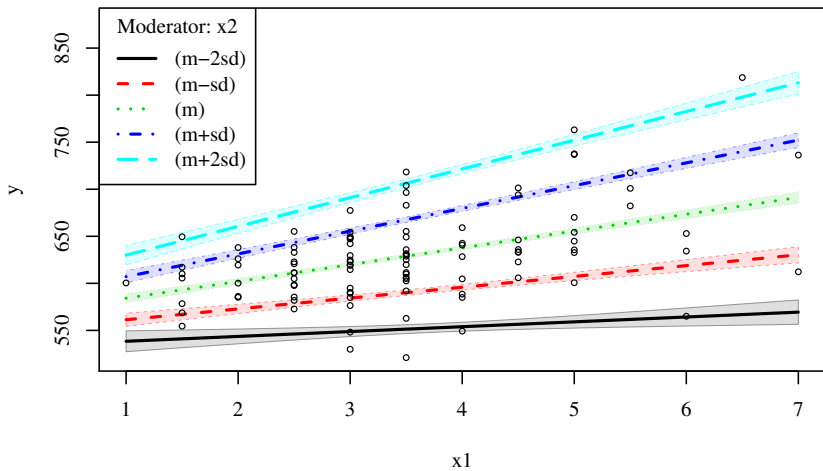


Want a lot of lines?  $n = 5$

```
ps8 <- plotSlopes(mod2, plotx = "x1", modx =  
  "x2", modxVals = "std.dev.", n = 5,  
  interval = "confidence")
```



## 5 lines



## Conclusion about plotSlopes

- If you don't want a plot, but rather just the newdata matrix and the predicted values, please look at `newdata()` and `predictOMatic()`.
- `plotCurves()` can do all of that stuff, and it works with nonlinear models and `glm`
- Have studied extension to other regression packages.
  - package writers are inconsistent, don't provide `predict` methods.
  - Conf. Intervals for `glm` objects controversial



## Analyzing Interaction effects

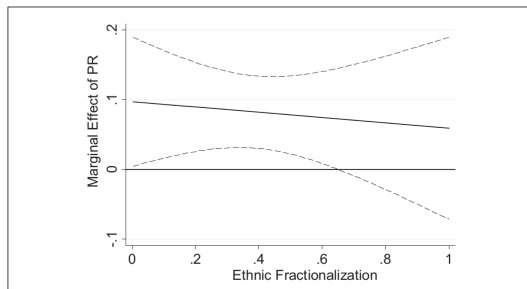
Selway & Templeman (2012) “Myth of Consociationalism?”

*Comparative Political Studies*

Model has PR\*EthnicFractionalization

Selway and Templeman

1559



**Figure 1.** Marginal effect of PR on *Political\_Deaths* across levels of ethnic fractionalization

The “marginal effect” is the slope  $(\hat{\beta}_{PR} + \hat{\beta}_{EF \cdot PR} EF_i) PR_i$



## testSlopes

- `plotSlopes()` creates an output object that allows a 'simple-slopes' analysis of statistical significance.
- If `modx` is
  - categorical**: simply calculates the slope of the relationship and tests whether it is different from 0
  - numeric**: calculates a Johnson-Neyman analysis: for which values of `modx` would the slope of `plotx` be different from 0?
- J-N: if the fitted model is  $\hat{y}_i = \hat{\beta}_0 + (\hat{\beta}_1 + \hat{\beta}_3 x_{2i}) x_{1i}$ , for which values of  $x_{2i}$  is  $(\hat{\beta}_1 + \hat{\beta}_3 x_{2i})$  statistically significantly different from 0?





# testSlopes

```
ps5ts <- testSlopes(ps5)
```

Values of x2 OUTSIDE this interval:

lo hi

42.79481 45.87360

cause the slope of  $(b1 + b2 * x2) * x1$  to be  
statistically significant



## A method for testSlopes objects (plot.testSlopes)

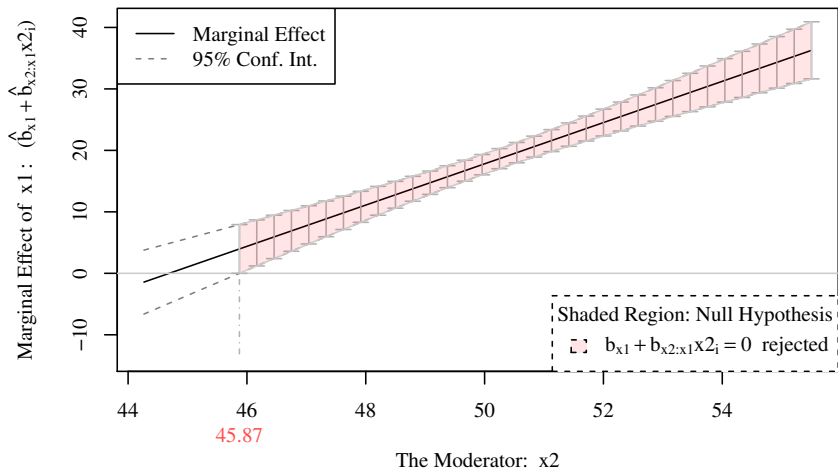
```
plot(ps5ts)
```

rockchalk is an S3 type R package.

If you are uncertain about the significance of S3 and the term “method”, I strongly recommend you get a copy of Friedrich Leisch, “Creating R Packages: A Tutorial” (available in CRAN contributed documentation) which has many excellent insights!



## plot of a testSlopes object



Note: intended verbosity of labels & legend

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## Mean-Center, Residual-Centered Regressions

- Start with
 
$$\text{lm}(y \sim x1 * x2 + x3, \text{data} = \text{dat})$$
- which implies
 
$$\text{lm}(y \sim x1 + x2 + x1:x2 + x3, \text{data} = \text{dat})$$
- Should it matter if we replace  $x1$  with
  - mean centered values,  $x1c = (x1 - \text{mean}(x1))$  by fitting
 
$$\text{lm}(y \sim x1c + x2 + x1c:x2 + x3, \text{data} = \text{dat})$$
- Or if we replace  $x1:x2$  by with the
  - “residual centered” value of the interaction term, which is the residual from this regression?
 
$$\text{lm}((x1*x2) \sim x1 + x2, \text{data} = \text{dat})$$



## Several authorities say those changes may be important

- Cohen, Cohen, Aichen & West (2002) strongly endorse mean-centering
- Little, T. D., Bovaird, J. A., & Widaman, K. F. (2006). On the Merits of Orthogonalizing Powered and Product Terms: Implications for Modeling Interactions Among Latent Variables. *Structural Equation Modeling*, 13(4), 497-519.



## 3 ease of use functions in rockchalk

- `standardize()` calculates centered & scaled values of all variables and re-fits the model.
- `meanCenter()` adjusts predictors by subtracting observed means
- `residualCenter()` calculates one variant of orthogonal regression
- rockchalk supplies `print()`, `predict()` and `summary()` methods for these functions



## Mean-Center

- Fit some big multiple regression

```
m1 <- lm(someDV ~ x1 + x2 + x3 * x4, data =  
  dat)
```

- Center only the interactive predictors

```
m1 <- lm(someDV ~ x1 + x2 + x3c*x4c, data =  
  dat)  
m1mc <- meanCenter(m1)
```

ends up fitting

```
lm(someDV ~ x1 + x2 + x3c + x4c + x3c:x4c,  
  data = dat)
```





## Mean-Center

- Center all predictors

```
m1mc2 <- meanCenter(m1, centerOnlyInteractors  
= FALSE)
```

ends up fitting

```
lm(someDV ~ x1c + x2c + x3c + x4c + x3c:x4c,  
data = dat)
```

- Center also the DV

```
m1mc3 <- meanCenter(m1, centerDV= TRUE,  
centerOnlyInteractors = FALSE)
```

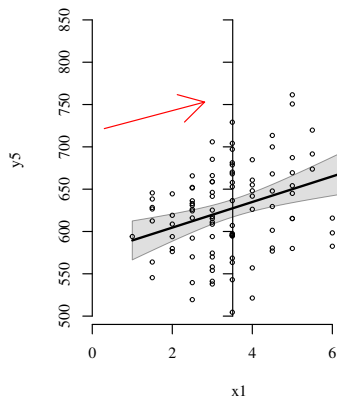
ends up fitting

```
lm(someDVc ~ x1c + x2c + x3c + x4c + x3c:x4c,  
data = dat)
```



## Why this is fool's gold

- Changing a predictor column from  $X_i$  to  $X_i - 5$  cannot improve statistical precision.
- It simply re-positions the Y axis.
  - Slope same, standard error of slope same
  - Intercept is “bigger”
  - Predicted value at Y axis is more precise, due to hour-glass shape of CI





## I was not so sure about residual centering

- The `residualCenter()` function leaves the linear terms in the model unchanged, but re-constructs interactive variables, replacing `x3:x4` with the residual from `lm(x3*x4 ~ x3+x4)`, which I'm calling "`x3.X.x4`"

```
m1rc <- lm(someDV ~ x1 + x2 + x3 + x4 + x5 +
           x6 + x5.X.x6 , data = dat)
```

- This is one way to create truly orthogonal columns. Before introduction of QR decomposition, it might have actually been a good way to do so
- Requires some serious fancy coding to make interactions like `x3*x4*x5` work correctly (see also `predict.mcreg()`)



# Alternatives seem better, but they are not actually different

- The predicted values are identical
- See the rockchalk vignette, which gives a full argument.
- In directory with this presentation, find the small example file `curve-example-1.R`

# Outline

- 1 Introduction
- 2 Data
- 3 Outreg
- 4 Plots
  - Categorical modx
  - Numeric moderator
- 5 Free Lunch
- 6 Conclusions**
- 7 Guessing

## Other functions worth mentioning

- mcDiagnose: splattering of multicollinearity diagnostics
- getDeltaRsquare, getPartialCor: partial and semi-partial correlations
- See the rockchalk vignette, which gives a full argument.
- In directory with this presentation, find the small example file curve-example-1.R

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# What makes package building easier?

roxygen2 (Hadley Wickham, Peter Danenberg, Manuel Eugster).

**Usual R development:** one writes R files, and documentation files in a separate directory. Very inconvenient to keep documents in sync with R code.

**roxygen2 approach:** put documentation in the R files, use functions to extract & format the documents.





## Am I competing with "car", "rms", "memisc", "texreg", etc?

- No. "car" and "rms" are established industry leading packages that support widely sold textbooks. Those authors are "up there", I'm "down here."
- No.
  - I'm filling in perceived gaps to create convenience
- Yes. Perhaps I think their
  - jargon is difficult (tough for me  $\implies$  impossible for students)
  - their functions are clumsy, or
  - I think their source code is not clear

