Extension and Customization

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1 Unsupported models: modelsummary_list

The simplest way to summarize an unsupported model is to create a modelsummary_list object. This approach is super flexible, but it requires manual intervention, and it can become tedious if you need to summarize many models. The next section shows how to add formal support for an unsupported model type.

A modelsummary_list is a list with two element that conform to the broom package specification: tidy and glance. tidy is a data.frame with at least three columns: term, estimate, and std.error. glance is a data.frame with only a single row, and where each column will be displayed at the bottom of the table in the goodness-of-fit section. Finally, we wrap those two elements in a list and assign it a modelsummary_list class:

```
ti <- data.frame(
  term = c("coef1", "coef2", "coef3"),
  estimate = 1:3,
  std.error = c(pi, exp(1), sqrt(2)))

gl <- data.frame(
  stat1 = "blah",
  stat2 = "blah blah")

mod <- list(
  tidy = ti,
  glance = gl)
class(mod) <- "modelsummary_list"

modelsummary(mod)</pre>
```

	(1)	
coef1	1.000	
	(3.142)	
coef2	2.000	
	(2.718)	
coef3	3.000	
	(1.414)	
stat1	blah	
stat2	blah blah	

2 Unsupported models: glance and tidy

modelsummary relies on two functions from the broom package to extract model information: tidy and glance. If broom doesn't support the type of model you are trying to summarize, modelsummary won't support it out of the box. Thankfully, it is extremely easy to add support for most models using custom methods.

For example, models produced by the MCMCglmm package are not currently supported by broom. To add support, you simply need to create a tidy and a glance method:

```
# load packages and data
library(modelsummary)
library(MCMCglmm)
data(PlodiaPO)
# add custom functions to extract estimates (tidy) and goodness-of-fit (glance) information
tidy.MCMCglmm <- function(x, ...) {</pre>
    s <- summary(x, ...)
    ret <- data.frame(</pre>
      term = row.names(s$solutions),
      estimate = s$solutions[, 1],
      conf.low = s$solutions[, 2],
      conf.high = s$solutions[, 3])
    ret
glance.MCMCglmm <- function(x, ...) {</pre>
    ret <- data.frame(</pre>
      dic = x\$DIC,
        = nrow(x$X))
    ret
}
# estimate a simple model
model <- MCMCglmm(PO ~ 1 + plate, random = ~ FSfamily, data = PlodiaPO, verbose=FALSE, pr=TR
# summarize the model
modelsummary(model, statistic = 'conf.int')
```

Three important things to note.

First, the methods are named tidy.MCMCglmm and glance.MCMCglmm because the model object I am trying to summarize is of class MCMCglmm. You can find the class of a model by running: class(model).

Second, both of the methods include the ellipsis . . . argument.

Third, in the example above we used the statistic = 'conf.int' argument. This is because the tidy method produces conf.low and conf.high columns. In most cases, users will define std.error column in their custom tidy methods, so the statistic argument will need to be adjusted.

If you create new tidy and glance methods, please consider contributing them to broom so that the rest of the community can benefit from your work: https://github.com/tidymodels/broom

3 Modifying information: tidy_custom and glance_custom

Users may want to include more information than is made available by the default extractor function. For example, models produced by the MASS::polr do not produce p values by default, which means that we cannot use the stars=TRUE argument in modelsummary. However, it is possible to extract this information by using the lmtest::coeftest function. To include such custom information, we will define new glance_custom and tidy_custom methods.

We begin by estimating a model with the MASS::polr:

```
library(MASS)
mod_ordinal <- polr(as.ordered(gear) ~ mpg + drat, data = mtcars)
get_estimates(mod_ordinal)</pre>
```

```
estimate std.error conf.level
                                           conf.low conf.high
                                                                statistic df.error
 term
                                                                                        р.
  3|4 13.962948761 4.04107300
                                    0.95 5.6851860 22.2407116
                                                               3.45525774
                                                                                28 0.00177
  4|5 16.898937342 4.39497069
                                    0.95 7.8962480 25.9016267
                                                                                28 0.00063
                                                               3.84506258
  mpg -0.008646682 0.09034201
                                    0.95 -0.1916706 0.1708667 -0.09571053
                                                                                28 0.92443
      3.949431923 1.30665144
                                    0.95 1.6191505 6.8457246
                                                               3.02255965
                                                                                28 0.00531
4 drat
```

The get_estimates function shows that our default extractor does *not* produce a p.value column. As a result, setting stars=TRUE in modelsummary will produce an error.

We know that the MASS::polr produces an object of class polr:

```
class(mod_ordinal)
```

```
[1] "polr"
```

To extract more (custom) information from a model of this class, we thus define a method called tidy_custom.polr which returns a data.frame with two columns: term and p.value:

```
tidy_custom.polr <- function(x, ...) {
    s <- lmtest::coeftest(x)
    out <- data.frame(
        term = row.names(s),
        p.value = s[, "Pr(>|t|)"])
    out
}
```

When this method is defined, modelsummary can automatically extract p values from all models of this class, and will now work properly with stars=TRUE:

```
modelsummary(mod_ordinal, stars = TRUE)
```

	(1)	
3 4	13.963**	
	(4.041)	
4 5	16.899***	
	(4.395)	
mpg	-0.009	
	(0.090)	
drat	3.949**	
	(1.307)	
Num.Obs.	32	
AIC	51.1	
BIC	57.0	
RMSE	3.44	
+ p < 0.1, * p < 0. ** p < 0.01, *** p		

4 New information: tidy_custom and glance_custom

Sometimes users will want to include information that is not supplied by those functions. A pretty easy way to include extra information is to define new glance_custom and tidy_custom methods. To illustrate, we estimate two linear regression models using the lm function:

```
library(modelsummary)

mod <- list()
mod[[1]] <- lm(hp ~ mpg + drat, mtcars)
mod[[2]] <- lm(wt ~ mpg + drat + am, mtcars)</pre>
```

In R, the 1m function produces models of class "lm":

```
class(mod[[1]])
```

[1] "lm"

Let's say you would like to print the dependent variable for each model of this particular class. All you need to do is define a new method called <code>glance_custom.lm</code>. This method should return a data.frame (or tibble) with 1 row, and 1 column per piece of information you want to display. For example:

```
glance_custom.lm <- function(x, ...) {
    dv <- as.character(formula(x)[2])
    out <- data.frame("DV" = dv)
    return(out)
}</pre>
```

Now, let's customize the body of the table. The vcov argument already allows users to customize uncertainty estimates. But imagine you want to override the *coefficient estimates* of your "lm" models. Easy! All you need to do is define a tidy_custom.lm method which returns a data.frame (or tibble) with one column called "term" and one column called "estimate".

Here, we'll substitute estimates by an up/down-pointing triangles which represents their signs:

```
tidy_custom.lm <- function(x, ...) {
    s <- summary(x)$coefficients
    out <- data.frame(
        term = row.names(s),
        estimate = ifelse(s[,1] > 0, '', ''))
    return(out)
}
```

After you define the glance_custom and tidy_custom methods, modelsummary will automatically display your customized model information:

```
modelsummary(mod)
```

	(1)	(2)
(Intercept)		
	(55.415)	(0.728)
mpg		
	(1.792)	(0.019)
drat		
	(20.198)	(0.245)
am		
		(0.240)
Num.Obs.	32	32
R2	0.614	0.803
R2 Adj.	0.588	0.782
AIC	337.9	46.4
BIC	343.7	53.7
Log.Lik.	-164.940	-18.201
F	23.100	38.066
RMSE	41.91	0.43
DV	hp	wt

Note that you can define a std.error column in tidy_custom.lm to replace the uncertainty estimates instead of the coefficients.

5 Customization: New model class

An even more fundamental way to customize the output would be to completely bypass modelsummary's extractor functions by assigning a new class name to your model. For example,

```
# estimate a linear model
mod_custom <- lm(hp ~ mpg + drat, mtcars)

# assign it a new class
class(mod_custom) <- "custom"

# define tidy and glance methods</pre>
```

```
tidy.custom <- function(x, ...) {
   data.frame(
     term = names(coef(x)),
     estimate = letters[1:length(coef(x))],
     std.error = seq_along(coef(x))
   )
}

glance.custom <- function(x, ...) {
   data.frame(
     "Model" = "Custom",
     "nobs" = stats:::nobs.lm(x)
   )
}

# summarize
modelsummary(mod_custom)</pre>
```

	(1)
(Intercept)	a
	(1.000)
mpg	b
	(2.000)
drat	\mathbf{c}
	(3.000)
Num.Obs.	32
Model	Custom

Warning: When defining new tidy and glance methods, it is important to include an ellipsis argument (...).

Note that in the glance.custom() method, we called stats:::nobs.lm() instead of the default stats::nobs() method, because the latter know does not know where to dispatch models of our new "custom" class. Being more explicit solves the problem.

An alternative would be to set a new class that inherits from the previous one, and to use a global option to set broom as the default extractor function (otherwise modelsummary will use its standard lm extractors by inheritance):

```
options(modelsummary_get = "broom")
class(mod_custom) <- c("custom", "lm")</pre>
```

6 Customization: modelsummary_list

Another flexible way to customize model output is to use output = "modelsummary_list". With this output option, modelsummary() returns a list with two elements: tidy contains parameter estimates, standard errors, etc., and glance contains model statistics such as the AIC. For example,

```
mod <- lm(hp ~ mpg + drat, mtcars)
mod_list <- modelsummary(mod, output = "modelsummary_list")
mod_list$tidy</pre>
```

```
term
                estimate std.error statistic df.error
                                                            p.value s.value group conf.low c
                                                    29 2.359726e-05
                                                                       15.4
1 (Intercept) 278.515455 55.414866 5.0260061
                                                                                        NA
                                                   29 5.172030e-06
2
             -9.985499 1.791837 -5.5727709
                                                                       17.6
                                                                                        NA
3
              19.125752 20.197756 0.9469246
                                                    29 3.515013e-01
         drat
                                                                        1.5
                                                                                        NA
```

```
mod_list$glance
```

```
aic bic r.squared adj.r.squared rmse nobs F logLik
1 337.8809 343.7438 0.6143611 0.5877653 41.90687 32 23.09994 -164.9404
```

Both tidy and glance can now be customized, and the updated model can be passed back to modelsummary using modelsummary(mod_list). All information that is displayed in the table is contained in mod_list, so this pattern allows for very flexible adjustments of output tables.

A useful example for this pattern concerns mixed models using 1me4. Assume we want to compare the effect of using different degrees-of-freedom adjustments on the significance of the coefficients. The models have identical parameter estimates, standard errors, and model fit statistics - we only want to change the p-values. We use the parameters package to compute the adjusted p-values.

```
library("lme4")
mod <- lmer(mpg ~ drat + (1 | am), data = mtcars)
mod_list <- modelsummary(mod, output = "modelsummary_list", effects = "fixed")
# create a copy, where we'll change the p-values</pre>
```

	Wald	Kenward
(Intercept)	-5.159	-5.159
	$6.409 \ (0.428)$	$6.409 \ (0.680)$
drat	7.045	7.045
	1.736 (<0.001) ***	$1.736\ (0.086)\ +$
Num.Obs.	32	32
R2 Marg.	0.402	0.402
R2 Cond.	0.440	0.440
AIC	188.7	188.7
BIC	194.6	194.6
ICC	0.1	0.1
RMSE	4.28	4.28