

## Statistical Formula Notation in R

R functions, notably `lm()` for fitting linear regressions and `glm()` for fitting logistic regressions, use a convenient formula syntax to specify the form of the statistical model to be fit. The basic format of such a formula is

response variable  $\sim$  predictor variables

The tilde is read as “is modeled as a function of.” A basic regression analysis would be formulated as

$$Y \sim X$$

Therefore we might fit a linear model regressing  $Y$  on  $X$  as

```
fit <- lm(Y ~ X)
```

where  $X$  is the predictor variable and  $Y$  is the response variable. In the usual mathematical notation this corresponds to the linear regression model denoted

$$Y_i = \beta_0 + \beta_1 X_i + \epsilon_i.$$

Additional explanatory variables can be included using the “+” symbol. To add another predictor variable  $Z$ , the formula becomes

$$Y \sim X + Z$$

and the linear regression call becomes

```
fit <- lm(Y ~ X + Z)
```

yielding a multiple regression with two predictors. The corresponding mathematical notation would be

$$Y_i = \beta_0 + X_i \beta_1 + Z_i \beta_2 + \epsilon_i.$$

Importantly, the use of the “+” symbol in this context is different than its usual meaning; the R formula notation is just a short-hand for which variable to include in the statistical model

and how. The following table lists the meaning of these symbols when used in an R modeling formula.

Symbol	Example	Meaning
+	+X	include this variable
-	-X	delete this variable
:	X:Z	include the interaction between these variables
*	X*Y	include these variables and the interactions between them
	X   Z	conditioning: include x given z
^	(X + Z + W) ^ 3	include these variables and all interactions up to three way
I	I (X*Z)	as is: include a new variable consisting of these variables multiplied
1	X - 1	intercept: delete the intercept (regress through the origin)

There is usually more than one way to specify the same model; the notation is not unique. For example the following three formulae are all equivalent:

```
Y ~ X + Z + W + X:Z + X:W + Z:W + X:Z:W
Y ~ X * Z * W
Y ~ (X + Z + W) ^ 3
```

each corresponding to the model

$$Y_i = \beta_0 + X_i\beta_1 + Z_i\beta_2 + W_i\beta_3 + X_iZ_i\beta_4 + X_iW_i\beta_5 + Z_iW_i\beta_6 + X_iZ_iW_i\beta_7 + \epsilon_i.$$

Likewise, each of these models

```
Y ~ X + Z + W + X:Z + X:W + Z:W
Y ~ X * Z * W - X:Z:W
Y ~ (X + Z + W) ^ 2
```

corresponds to

$$Y_i = \beta_0 + X_i\beta_1 + Z_i\beta_2 + W_i\beta_3 + X_iZ_i\beta_4 + X_iW_i\beta_5 + Z_iW_i\beta_6 + \epsilon_i,$$

which differs from the previous model in that the three-way interaction has been omitted.

Finally, when using a data frame an additional time-saver is to use "." to indicate "include all variables". This is especially convenient when used in conjunction with the other symbols. Consider a data frame D which has columns Y, X, Z, and W. Then the function call

```
fit <- lm(Y ~ ., data = D)
```

is equivalent to

```
fit <- lm(Y ~ X + Z + W, data = D)
```

Similarly,

```
fit <- lm(Y ~ .-W, data = D)
```

is equivalent to

```
fit <- lm(Y ~ X + Z)
```

and

```
fit <- lm(Y ~ .*W, data = D)
```

is equivalent to

```
fit <- lm(Y ~ X + Z + W + X:W + Z:W)
```

Using this notation permits a data analyst to run a spate of regression specifications without having to reconfigure the columns of a spreadsheet each time.