

# Lecture 14: Other Models

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So far, we have seen how to use the function `lm_basic` to build linear regression models.

As I probably mentioned before, this is just a soft wrapper around the base R function `lm`, with the following tweaks:

- ▶ you are forced to supply a data frame
- ▶ the summary table produces confidence intervals rather than standard errors

I don't use this function in my own work, but find it easier for teaching.

One major benefit of the “real” `lm` is that there are many more options available for calling it, such as including sample weights.

There are many other functions in R that have similar calling mechanisms to `lm` but run different underlying models.

For example, `glm` fits generalized linear models. These can be used, amongst other things, for fitting a model to a binary response:

```
df <- data_frame(y = c(0,0,0,0,1,1,1,1), x = rnorm(8))  
model <- glm(y ~ x, data = df, family = binomial())
```

# glm()

```
summary(model)
```

```
##
```

```
## Call:
```

```
## glm(formula = y ~ x, family = binomial(), data = df)
```

```
##
```

```
## Deviance Residuals:
```

| ## | Min      | 1Q       | Median  | 3Q      | Max     |
|----|----------|----------|---------|---------|---------|
| ## | -1.28128 | -1.13899 | 0.01232 | 1.15536 | 1.24458 |

```
##
```

```
## Coefficients:
```

| ## |             | Estimate | Std. Error | z value | Pr(> z ) |
|----|-------------|----------|------------|---------|----------|
| ## | (Intercept) | -0.02157 | 0.71807    | -0.030  | 0.976    |
| ## | x           | 0.16454  | 0.88998    | 0.185   | 0.853    |

```
##
```

```
## (Dispersion parameter for binomial family taken to be 1)
```

```
##
```

In the **MASS** package (included with all standard R installations) is the `rlm` function for fitting robust linear regression:

```
library(MASS)
x <- rnorm(100)
y <- 1 + x + rnorm(100, sd = 0.2)
y[50] <- 1e4
model_lm <- lm(y ~ x)
model_rlm <- rlm(y ~ x)
```

# rlm()

```
summary(model_lm)
```

```
##
```

```
## Call:
```

```
## lm(formula = y ~ x)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
```

```
## -433.7 -184.2  -99.7   -15.0  9734.3
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept)    117.1      100.4   1.166   0.246
```

```
## x              -142.3      110.9  -1.283   0.203
```

```
##
```

```
## Residual standard error: 996.6 on 98 degrees of freedom
```

```
## Multiple R-squared:  0.01651,    Adjusted R-squared:  0.006479 7
```

# rlm()

```
summary(model_rlm)
```

```
##
```

```
## Call: rlm(formula = y ~ x)
```

```
## Residuals:
```

```
##           Min           1Q           Median           3Q           Max
## -4.573e-01 -1.245e-01 -2.746e-02  1.196e-01  1.000e+04
##
```

```
## Coefficients:
```

```
##           Value   Std. Error t value
## (Intercept)  0.9980   0.0227   44.0345
## x            0.9965   0.0250   39.8049
```

```
##
```

```
## Residual standard error: 0.1934 on 98 degrees of freedom
```



If you have a need for a specific model, you can usually find an R package that support it. In most cases, the model will roughly resemble calling `lm`.

Some common examples you may run into:

- ▶ `gam::gam` for generalized additive models
- ▶ `nls` for non-linear regression
- ▶ `lme4::lmer` for mixed effects models
- ▶ `quantreg::qr` for quantile regression
- ▶ `glmnet::glmnet` for the generalized elastic net
- ▶ `randomforest::randomforest` for random forest classifications
- ▶ `forecast::auto.arima` for modeling time series