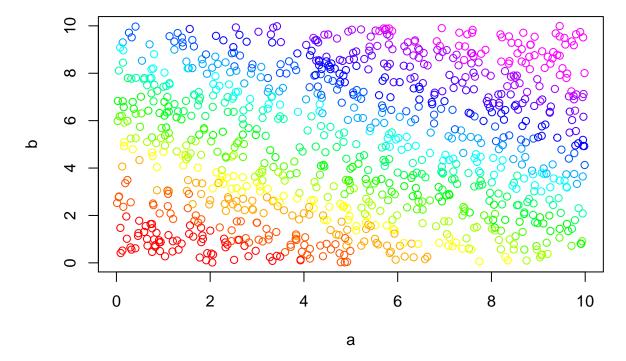
## Multicolinearity

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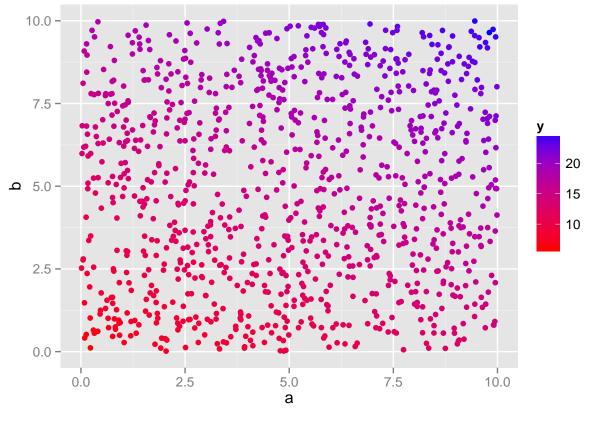
```
N <- 1e3  # 1e5
a <- runif(N, min=0, max=10)
b <- runif(N, min=0, max=10)
s <- 1.2
y <- 6 + 0.7 * a + 1.2 * b + rnorm(N, sd=0.2)
df1 <- data.frame(a, b, y)</pre>
```

To display three dimensions, we can use color. Here we break the y values into a series of ranges, and assign a color to each range. Colors are made by the rainbow function, which makes a series of hues spanning the spectrum.

```
df1$y_bucket <- cut(y, breaks=quantile(y, probs=0:16/16))
rbow <- rainbow(16, end=5/6)
with(df1, plot(x=a, y=b, col=rbow[y_bucket]))</pre>
```



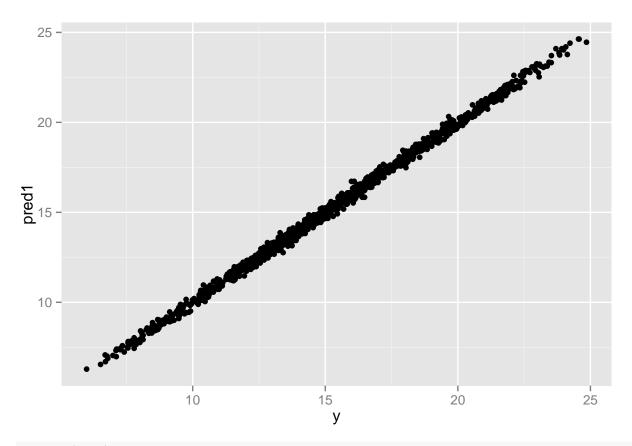
```
library(ggplot2)
ggplot(df1, aes(x=a, y=b, col=y)) +
   geom_point() +
   scale_colour_gradient(low="red", high="blue")
```



```
fit1 <- lm( y ~ a + b, data=df1)

df1$pred1 <- fit1$fitted

ggplot(df1, aes(x=y, y=pred1)) + geom_point()</pre>
```

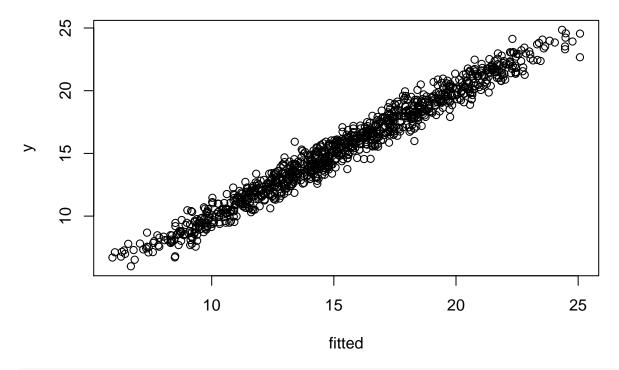


## summary(fit1)

```
##
## Call:
## lm(formula = y ~ a + b, data = df1)
##
## Residuals:
##
       \mathtt{Min}
                 1Q Median
                                   ЗQ
## -0.73273 -0.15768 -0.00722 0.15179 0.64761
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 5.987215 0.017352
                                   345.0 <2e-16 ***
             0.701205
                         0.002347
                                    298.8
                                            <2e-16 ***
              1.201904 0.002321
                                   517.9 <2e-16 ***
## b
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.2135 on 997 degrees of freedom
## Multiple R-squared: 0.9973, Adjusted R-squared: 0.9973
## F-statistic: 1.866e+05 on 2 and 997 DF, p-value: < 2.2e-16
# (I want the fit to be good, but the coefficients to be non-significant.
# Try more multicollinear columns:
```

Matrix version

```
num_a_cols <- 4</pre>
num_b_cols <- 4
X_signal <- matrix( c(rep(a, num_a_cols), rep(b, num_b_cols)), ncol=(num_a_cols + num_b_cols) )</pre>
X_noise <- matrix( rnorm( (num_a_cols + num_b_cols) * N), ncol=(num_a_cols + num_b_cols) )</pre>
X <- X_signal + X_noise</pre>
df2 <- cbind(data.frame(X), y)</pre>
fitA \leftarrow lm(y \sim ., data=df2)
summary(fitA)
##
## Call:
## lm(formula = y ~ ., data = df2)
##
## Residuals:
       Min
                10 Median
                                 3Q
                                        Max
## -2.4042 -0.4908 -0.0143 0.4939
                                     2.5290
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
                           0.05802 108.579 < 2e-16 ***
## (Intercept) 6.29979
## X1
                0.21362
                           0.02087 10.235 < 2e-16 ***
## X2
                                     7.728 2.66e-14 ***
                0.15282
                           0.01977
                                     7.823 1.32e-14 ***
## X3
                0.15720
                           0.02009
                0.15132
                           0.02017
                                     7.504 1.38e-13 ***
## X4
## X5
                0.31577
                           0.01959 16.119 < 2e-16 ***
## X6
                0.27937
                           0.02037 13.717 < 2e-16 ***
## X7
                0.28543
                           0.01989 14.352 < 2e-16 ***
## X8
                0.27701
                           0.02073 13.361 < 2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.7215 on 991 degrees of freedom
## Multiple R-squared: 0.9698, Adjusted R-squared: 0.9695
## F-statistic: 3973 on 8 and 991 DF, p-value: < 2.2e-16
df2$fitted <- fitA$fitted
plot(y ~ fitted, data=df2)
```



df2\$fitted <- fitA\$fitted