Galton Regression

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2025-08-28

Reading Galton's Data

```
DATA=read.csv('https://raw.githubusercontent.com/gdlc/STAT_COMP/master/DATA/GALTON.csv',header=T)
```

Adding the parental average (PA)

```
DATA$PA=(DATA$Father+DATA$Mother)/2
```

Regressions

```
fmMother=lm(Height~Mother,data=DATA)
fmFather=lm(Height~Father,data=DATA)
fmPA=lm(Height~PA,data=DATA)
fmPA2=lm(Height~PA+Gender,data=DATA)
summary(fmMother)
##
## Call:
## lm(formula = Height ~ Mother, data = DATA)
##
## Residuals:
##
      Min
               1Q Median
                               ЗQ
                                      Max
## -9.5474 -2.6346 -0.1079 2.8688 11.9526
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 46.69077
                          3.25874 14.328 < 2e-16 ***
## Mother
              0.31318
                          0.05082
                                    6.163 1.08e-09 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 3.511 on 896 degrees of freedom
## Multiple R-squared: 0.04066,
                                   Adjusted R-squared: 0.03959
## F-statistic: 37.98 on 1 and 896 DF, p-value: 1.079e-09
summary(fmFather)
##
## Call:
```

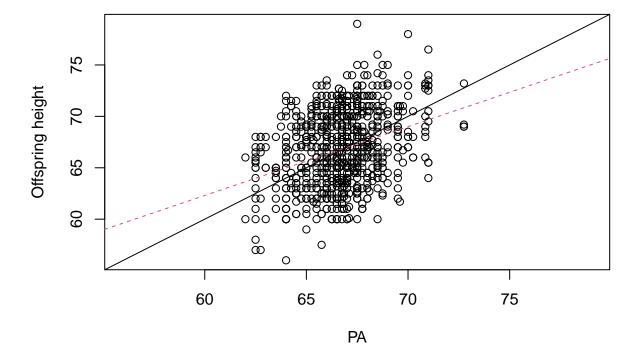
```
## lm(formula = Height ~ Father, data = DATA)
##
## Residuals:
##
                 1Q
                                   3Q
       Min
                     Median
                                           Max
## -10.2683 -2.6689 -0.2092
                               2.6342 11.9329
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 39.11039
                          3.22706 12.120
                                            <2e-16 ***
## Father
              0.39938
                          0.04658
                                  8.574
                                            <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 3.446 on 896 degrees of freedom
## Multiple R-squared: 0.07582,
                                   Adjusted R-squared:
## F-statistic: 73.51 on 1 and 896 DF, p-value: < 2.2e-16
summary(fmPA)
##
## Call:
## lm(formula = Height ~ PA, data = DATA)
## Residuals:
      Min
               1Q Median
                               30
                                      Max
## -8.9814 -2.6604 -0.1642 2.7795 11.6762
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 22.1488
                           4.3076
                                    5.142 3.34e-07 ***
## PA
                0.6693
                           0.0646 10.360 < 2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 3.388 on 896 degrees of freedom
## Multiple R-squared: 0.107, Adjusted R-squared: 0.106
## F-statistic: 107.3 on 1 and 896 DF, p-value: < 2.2e-16
```

Beyond the In-class assignment

What follows was not part of the assignement.

Galton's plot

```
tmp=range(c(DATA$Height,DATA$PA))
plot(Height~PA,data=DATA,xlim=tmp,ylim=tmp,xlab='PA',ylab='Offspring height');
abline(0,1);
abline(a =coef(fmPA)[1],b=coef(fmPA)[2],col=2,lty=2)
```



The slope of the regressions and trait heritability

Fisher's infinitesimal model can be described as follows

$$Y_i = G_i + E_i$$

Where Y_i is a phenotype measurement on the ith subject, G_i is a genetic (random) effect (capturing the collective effects of all the loci affecting the trait), and E_i is an environmental effect, capturing all the non-genetic effects.

Because of the Central Limit Theorem, if the number of genes influencing the trait is large, $= G_i$ follows (approximately) a normal distribution.

Using the above model we can decompose the phenotypic variance $V_p = Var(Y_i)$ as follows

$$Var(Y_i) = Var(G_i) + Var(E_i) + 2Cov(G_i, E_i)$$

In absence of genetic-by-environmental covaraince (i.e., if the environmental effects do not depend on individuals' genotypes), we have

$$Var(Y_i) = Var(G_i) + Var(E_i)$$

Or

$$V_P = V_G + V_E$$

The trait heritability is the proportion of variance of the phenotypes explained by genetic factors, that ids

$$h^2 = \frac{V_G}{V_G + V_E}$$

Slopes

The slope on a linear regression of Y on X is b = Cov(X, Y)/Var(X).

In the Mother's regression we have

$$b_M = Cov(Y_o, Y_m)/Var(Y_m)$$

For the offspring equation we can write

$$Y_o = (0.5 \times G_f + 0.5 \times G_m + \Psi) + E_o$$

where, G_f and G_m are the genetic values of the father and mother, the genetic value of the offspring is

$$G_o = (0.5 \times G_f + 0.5 \times G_m + \Psi)$$

the sum of half of the genetic value of the parents plus a term, Ψ , called mendelian sampling, that captures the randomness resulting from the sampling of alleles at the meiosis.

Using the above, and using the equation for mothers' phenotype, $Y_m = G_m + E_m$, we have

$$Cov(Y_o, Y_m) = 0.25 \times V_q + 0.25 \times V_q = 0.5 \times V_q$$

Therefore, the slope of the regression is

$$b = \frac{0.5 \times V_g}{V_p} = 0.5 \times h^2$$

The heritability of height is ~0.8, if we adjust by sex, the estimated slope is $(b = 0.5 \times h^2)$, the estimated slope is

```
fmMother=lm(Height~Mother+Gender,data=DATA)
summary(fmMother)
```

```
##
## Call:
## lm(formula = Height ~ Mother + Gender, data = DATA)
##
## Residuals:
##
       Min
                1Q
                    Median
                                 3Q
                                        Max
##
  -9.4036 -1.6024
                    0.1528
                            1.5890
                                     9.4199
##
##
  Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 41.44952
                            2.20949
                                      18.76
                                              <2e-16 ***
## Mother
                0.35314
                            0.03439
                                      10.27
                                              <2e-16 ***
## GenderM
                5.17669
                            0.15867
                                      32.62
                                              <2e-16 ***
##
## Signif. codes:
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2.374 on 895 degrees of freedom
## Multiple R-squared: 0.5618, Adjusted R-squared: 0.5608
## F-statistic: 573.7 on 2 and 895 DF, p-value: < 2.2e-16
```

```
confint(fmMother)
                    2.5 %
                               97.5 %
## (Intercept) 37.1131324 45.7859147
## Mother
                0.2856502 0.4206241
## GenderM
                4.8652824 5.4881075
Under the model we discussed, the slope for Fathers' regression is the same (b = 0.5 \times h^2), from the data we
fmMother=lm(Height~Father+Gender,data=DATA)
summary(fmMother)
##
## Call:
## lm(formula = Height ~ Father + Gender, data = DATA)
##
## Residuals:
##
       Min
                1Q Median
                                 3Q
                                        Max
## -9.3708 -1.4808 0.0192 1.5616 9.4153
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 34.46113
                            2.13628
                                      16.13
                                              <2e-16 ***
                                      13.90
## Father
                0.42782
                            0.03079
                                              <2e-16 ***
                5.17604
                            0.15211
                                      34.03
## GenderM
                                              <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.277 on 895 degrees of freedom
## Multiple R-squared: 0.5971, Adjusted R-squared: 0.5962
## F-statistic: 663.2 on 2 and 895 DF, p-value: < 2.2e-16
confint(fmMother)
##
                    2.5 %
                               97.5 %
## (Intercept) 30.2684263 38.6538353
## Father
                0.3674023 0.4882411
## GenderM
                4.8775165 5.4745684
```

Question: How do we calculate the slope for the regression involving PA?

In both cases, 0.4 $(1/2 \text{ of } h_{height}^2)$ is within the CI.