Finding correlation for acmPRS

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Goals:

- Find Var(acmPRS)
- Find Cov(Y, acmPRS)

0.1 Finding Var(acmPRS):

Define

$$\hat{S}_i = \sum_{j=1}^{m_i} \hat{\beta} \mathbf{G_j}$$

$$\hat{S}_{acm} = \sum_{i=1}^{c} n_i r_i \sqrt{\frac{h_i}{h_1}} \hat{S}_i$$

Theorem 1. $E[\hat{S}_{acm}] = 0$

Proof.

$$\begin{split} E[\hat{S}_{acm}] &= E[\sum_{i=1}^{c} n_i r_i \sqrt{\frac{h_i}{h_1}} \hat{S}_i] \\ &= \sum_{i=1}^{c} E[n_i r_i \sqrt{\frac{h_i}{h_1}} \hat{S}_i] \\ &= \sum_{i=1}^{c} n_i r_i \sqrt{\frac{h_i}{h_1}} E[\hat{S}_i] \end{split}$$

$$E[\hat{S}_i] = 0$$

$$E[\hat{S}_{acm}] = 0$$

Theorem 2. $Var(\hat{S}_{acm}) = \sum_{i=1}^{c} \frac{m_i n_i^2 r_i^2 h_i}{h_1} Var(\hat{\beta}_i) + \sum_{i \neq j}^{c} n_i n_j r_i r_j \sqrt{\frac{h_i h_j}{h_1^2}} Cov(\hat{S}_i, \hat{S}_j)$

Proof.

$$Var(\sum_{i=1}^{c} n_{i}r_{i}\sqrt{\frac{h_{i}}{h_{1}}}\hat{S}_{i}) = \sum_{i=1}^{c} Var\left(n_{i}r_{i}\sqrt{\frac{h_{i}}{h_{1}}}\hat{S}_{i}\right) + \sum_{i\neq j} Cov\left(n_{i}r_{i}\sqrt{\frac{h_{i}}{h_{1}}}\hat{S}_{i}, n_{j}r_{j}\sqrt{\frac{h_{j}}{h_{1}}}\hat{S}_{j}\right)$$

$$\sum_{i=1}^{c} Var\left(n_{i}r_{i}\sqrt{\frac{h_{i}}{h_{1}}}\hat{S}_{i}\right) = \sum_{i=1}^{c} n_{i}r_{i}\sqrt{\frac{h_{i}}{h_{1}}}Var(\hat{S}_{i})$$

$$= \sum_{i=1}^{c} n_{i}r_{i}\sqrt{\frac{h_{i}}{h_{1}}}m_{i}Var(\hat{\beta}_{i})$$

$$(1)$$

$$\sum_{i \neq j} Cov \left(n_i r_i \sqrt{\frac{h_i}{h_1}} \hat{S}_i, n_j r_j \sqrt{\frac{h_j}{h_1}} \hat{S}_j \right) = \sum_{i \neq j} E \left[\left(n_i r_i \sqrt{\frac{h_i}{h_1}} \hat{S}_i \right), \left(n_j r_j \sqrt{\frac{h_j}{h_1}} \hat{S}_j \right) \right] + E \left[\sum_{i=1}^c n_i r_i \sqrt{\frac{h_i}{h_1}} \hat{S}_i \right]^c e^{\sum_{j=1}^c n_j r_j \sqrt{\frac{h_j}{h_1}} \hat{S}_j} \right]^{-0}$$

$$= \sum_{i \neq j} n_i n_j r_i r_j \sqrt{\frac{h_i h_j}{h_1^2}} E[\hat{S}_i \hat{S}_j]$$

$$= \sum_{i \neq j} n_i n_j r_i r_j \sqrt{\frac{h_i h_j}{h_1^2}} Cov \left(\hat{S}_i, \hat{S}_j \right)$$
(2)

The above is because

$$Cov(\hat{S}_i, \hat{S}_j) = E\left[\hat{S}_i \hat{S}_j\right] + E[\hat{S}_i] + E[\hat{S}_j]^0$$

$$= E\left[\hat{S}_i \hat{S}_j\right]$$

Combining (1) and (2):

$$Var(\hat{S}_{acm}) = \sum_{i=1}^{c} \frac{m_{i}n_{i}^{2}r_{i}^{2}h_{i}}{h_{1}} Var(\hat{\beta}_{i}) + \sum_{i \neq j}^{c} n_{i}n_{j}r_{i}r_{j} \sqrt{\frac{h_{i}h_{j}}{h_{1}^{2}}} Cov(\hat{S}_{i}, \hat{S}_{j})]$$

Theorem 3. $Cov(Y, acmPRS) = \frac{1}{c} \sum_{i=1}^{c} n_i r_i \sqrt{\frac{h_i}{h_1}} m_i Cov(\hat{\beta}_i, \beta_Y)$

Proof.

$$\begin{aligned} \operatorname{Cov}(Y, \operatorname{acmPRS}) &= E[Y \operatorname{acmPRS}] + E[Y] + \underbrace{E[\operatorname{acmPRS}]}^0 \\ &= E\left[\left(\sum_{j=1}^m \beta_{jY} G_i \right) \left(\sum_{i=1}^c n_i r_i \sqrt{\frac{h_i}{h_1}} \sum_{j=1}^m \hat{\beta}_{ji} G_j \right) \right] \end{aligned}$$

I'm really not sure if I'm allowed to do this.

$$E\left[\left(\sum_{j=1}^{m}\beta_{jY}G_{i}\right)\left(\sum_{i=1}^{c}n_{i}r_{i}\sqrt{\frac{h_{i}}{h_{1}}}\sum_{j=1}^{m}\hat{\beta}_{ji}G_{j}\right)\right] = E\left[\sum_{i=1}^{c}n_{i}r_{i}\sqrt{\frac{h_{i}}{h_{1}}}\left[\sum_{j=1}^{m_{i}}\beta_{j}G_{j}\right]\frac{1}{c}\left[\sum_{j=1}^{m_{i}}\beta_{jY}G_{j}\right]\right]$$

Here I've brought the Y score inside the summation with a $\frac{1}{c}$ coefficient.

$$=\frac{1}{c}\sum_{i=1}^{c}\left[n_{i}r_{i}\sqrt{\frac{h_{i}}{h_{1}}}E\left[\sum^{m_{i}}\beta_{j}G_{j}\sum^{m_{Y}}\beta_{jY}G_{j}\right]\right]$$

And since by equation (7) in power and predictive accuracy: $E\left[\sum^{m_i} \beta_j G_j \sum^{m_Y} \beta_{jY} G_j\right] = m\text{Cov}(\hat{\beta}_i, \beta_Y)$

Then

$$Cov(Y, acmPRS) = \frac{1}{c} \sum_{i=1}^{c} n_i r_i \sqrt{\frac{h_i}{h_1}} m_i Cov(\hat{\beta}_i, \beta_Y)$$

$$\textbf{Theorem 4.} \ \ R^2_{Y,acmPRS} = \frac{\frac{1}{c^2} \left(\sum_{i=1}^c n_i r_i \sqrt{\frac{h_i}{h_1}} m_i Cov(\hat{\beta}_i, \beta_Y) \right)^2}{\left(\sum_{i=1}^c \frac{m_i n_i^2 r_i^2 h_i}{h_1} Var(\hat{\beta}_i) + \sum_{i \neq j}^c n_i n_j r_i r_j \sqrt{\frac{h_i h_j}{h_1^2} Cov(\hat{S}_i, \hat{S}_j)]} \right) \left(\sigma_g^2 + \sigma_e^2 \right)}$$

Proof.

$$R_{X,Y}^2 = \frac{Cov(X,Y)^2}{Var(X)Var(Y)}$$

Sub in equations in theorems (2) and (3).