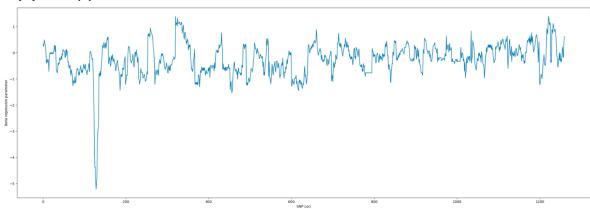
(a) Univariate regression

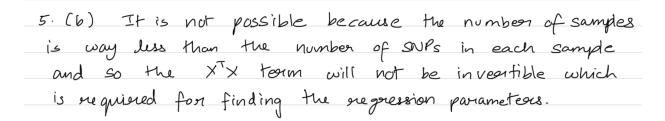
• [10 pts] Perform a univariate regression, using the LEU2 expression levels as output and each SNP as input. Obtain $\beta = [\beta_1, \dots, \beta_J]$, where J = 1,260 and β_i is the regression parameter estimated by regressing LEU2 expression levels on SNP i. Plot β_i 's across the genome.

5.(a) See python file for code. Plot attached below:



(b) Multivariate regression

• [10 pts] Can you obtain an estimate of the regression coefficients for a multivariate regression model using all SNPs as inputs and *LEU*2 expression as output? Explain why this is not possible. [Hint: you do not need to write a code to answer this question.]

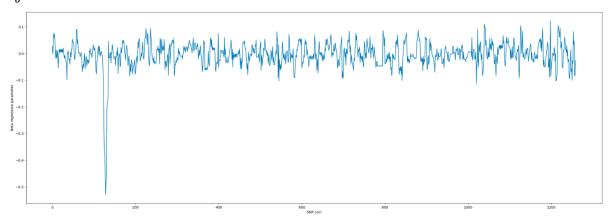


(c) Ridge regression.

• [10 pts] Perform a ridge regression, assuming a prior distribution $\beta_i \sim N(0, \sigma_0^2)$ for $i=1,\ldots,J,\,J=1,260$, with $\sigma_0^2=5.0$. Your multivariate regression model should predict LEU2 expression levels given all SNPs. Plot your estimated regression coefficients $\beta=[\beta_1,\ldots,\beta_J]$.

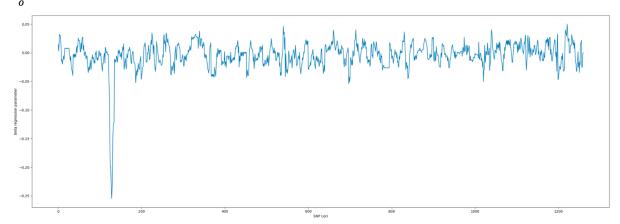
5.(c) See python file for code. Plots attached below:

$$\sigma_{o}^{2} = 5.0$$



• [10 pts] Repeat the analysis above with a prior distribution $\beta_i \sim N(0, \sigma_0^2)$ for $i = 1, \ldots, J$, J = 1, 260, with $\sigma_0^2 = 0.005$. Plot your estimated regression coefficients $\beta = [\beta_1, \ldots, \beta_J]$.

$$\sigma_o^2 = 0.005$$



• [10 pts] Explain the different effects of the two prior distributions above on the regression parameter estimates. Which SNP has the strongest influence on the LEU2 expression?

The resulting slope using a lesser σ_0^2 (0.005) is less steeper than using the higher σ_0^2 (5.0). This means that the predictions for the parameters becomes less sensitive to the SNP loci as the σ_0^2 decreases and s increases. Small σ_0^2 means large S and a steronger parior belief and it means steronger pull of estimates toward zero.

SNP no. 128 (0-based index) has highest influence. It is called YDR085C.