

## Question 4

### What is Parametric Bootstrap?

We can sample from the distribution  $F$  but instead we obtain a sample using an estimate  $\hat{F}$ . We can estimate the distribution function  $F(x)$  using a parametric model  $F(x; \theta)$  which is indexed by some parameters.

For a given sample  $S$  and parametric model  $F(x; \theta)$  we obtain an estimate  $\hat{\theta}$  based on the sample. We then generate  $B$  bootstrap samples  $S_1^*, \dots, S_B^*$  using  $F(x; \hat{\theta})$ . Here we generate samples from the model and not through sampling with replacement.

### Parametric Bootstrap for Regression

We will now apply the parametric bootstrap in the context of regression.

- The assumed regression model is

$$Y_i = \alpha + \beta(x_i - \bar{x}) + R_i$$

where  $R_i \approx_{iid} G(0, \sigma)$  (iid stands for independent and identically distributed random variables)

- We fit the model to obtain the estimates  $\hat{\alpha}, \hat{\beta}$  and  $\hat{\sigma}$
- After obtaining the above values  $\hat{\alpha}, \hat{\beta}$  and  $\hat{\sigma}$  we fit them in the above regression model

Now in order to obtain bootstrap samples, we will generate  $R_i^*$  from  $G(0, \hat{\sigma})$  and set  $y_i^* = \hat{\alpha} + \hat{\beta}(x_i - \bar{x}) + R_i^*$  where  $x_i$  are fixed.

We then obtain the bootstrap sample as:

$$S_b^* = \{(x_1, y_1^*), (x_2, y_2^*), \dots, (x_n, y_n^*)\}$$

and as we saw  $x_i$  is fixed hence in this case  $x_1, x_2, \dots, x_n$  is fixed and  $y_1^*, y_2^*, \dots, y_n^*$  are different.

For each bootstrap sample  $S_b^*$  we estimate the parameters to get the bootstrap replicates  $\hat{\alpha}_b^*, \hat{\beta}_b^*$  and  $\hat{\sigma}_b^*$

- The parametric bootstrap motivates another way to re-sample data i.e. sampling the errors