

1. In the multiple linear regression, we assume that...

- 10/13 ☒ A The number of observations is much larger than the number of variables ($n \gg p$)
- 3/13 ☐ B The number of observations is slightly larger than the number of variables ($n > p$)
- 0/13 ☐ C The number of observations equals than the number of variables ($n = p$)
- 0/13 ☐ D The number of observations is less than the number of variables ($n < p$)
- 0/13 ☐ E It is not important
- 0/13 ☐ F I do not know

2. The way of solving the problem of a large number of variables is...

- 10/13 ☒ A Subset Selection & Shrinkage (Regularization)
- 1/13 ☐ B Shrinkage (Regularization) & Maximum Likelihood estimation
- 2/13 ☐ C Dimension Reduction & OLS estimation
- 0/13 ☐ D I do not know
- 0/13 ☐ E The absence of the right answer

3. The bias of an estimator (e.g. \hat{z}) equals

- 10/13 ☒ A $E(\hat{z}) - z$
- 1/13 ☐ B $E(\hat{z}^2) - [E(z)]^2$
- 2/13 ☐ C $[E(\hat{z}^2) - E(z)]^2$
- 0/13 ☐ D $E(\hat{z}^2)$
- 0/13 ☐ E I do not know

4. The main idea of regularization is

- 9/13 ☒ A To introduce a small amount of bias in order to have less variance.
- 1/13 ☐ B To introduce a small amount of variance in order to have less bias.
- 3/13 ☐ C To introduce a small amount of variance and bias in order to have less bias.
- 0/13 ☐ D I do not know

5. With which function we can show regularization in R

- 11/13 ☒ A glmnet()
- 0/13 ☐ B regular()
- 0/13 ☐ C lm()
- 2/13 ☐ D glm()
- 0/13 ☐ E I do not know

6. How the tune of any parametr can be made

- 12/13 ☒ A using Cross validation
- 0/13 ☐ B It is impossible
- 0/13 ☐ C I do not now
- 1/13 ☐ D using larger sample
- 0/13 ☐ E only having population

7. Elastic Net is

- 12/13 ☒ A the combination of L1 and L2 regularization
- 1/13 ☐ B the combination of L2 and L3 regularization
- 0/13 ☐ C is independent from other types of regularization
- 0/13 ☐ D I do not know
- 0/13 ☐ E not a type of regularization

8. Regularization is used only for

- 1/13 ☐ A Poisson Regression
- 6/13 ☐ B Linear Regression
- 0/13 ☐ C Logistic Regression
- 6/13 ☒ D any regression
- 0/13 ☐ E I do not know

9. Regularization can solve the problem of

- 0/13 ☐ A heteroscedasticity
- 13/13 ☒ B multicollinearity
- 0/13 ☐ C autocorrelation
- 0/13 ☐ D I do not know

10. As a result of regularization, we will have

- 10/13 ☒ A smaller slope than in case of OLS
- 1/13 ☐ B larger slope than in case of OLS
- 2/13 ☐ C the slope remains the same
- 0/13 ☐ D I do not know

11. The ridge coefficient estimates shrink towards zero

- 10/13 ☒ A when lambda increases
- 2/13 ☐ B when lambda decreases
- 1/13 ☐ C when lambda = 0
- 0/13 ☐ D I do not know

12. Which one can shrink the slope all the way to 0?

- 10/13 ☒ A Lasso
- 3/13 ☐ B Ridge
- 0/13 ☐ C Regression
- 0/13 ☐ D I do not know

13. When lambda = 0, we have

- 1/13 ☐ A Ridge
- 2/13 ☐ B Lasso
- 1/13 ☐ C EL
- 9/13 ☒ D Regression
- 0/13 ☐ E I do not know

14. When alpha = 0, we have

- 8/13 ☒ A Ridge
- 2/13 ☐ B Lasso
- 2/13 ☐ C EL
- 1/13 ☐ D Regression
- 0/13 ☐ E I do not know

15. Which function can help to perform cross-validation for regularization in R?

9/13 **A** ☒ cv.glmnet()

1/13 **B** ☐ cros_val()

3/13 **C** ☐ glmnet(method = "cv")

0/13 **D** ☐ I do not know

16. Why we use set.seed() in R?

10/13 **A** ☒ To have universal result

2/13 **B** ☐ To perform better result

1/13 **C** ☐ To have random models

0/13 **D** ☐ I do not know