

✔ Congratulations! You passed!

Grade received 90.90%

Latest Submission Grade 90.91%

To pass 80% or higher

Go to next item

1. Mean-centering the data helps MCMC Sampling by

1 / 1 point

- ☒ Reducing the correlation between the variables
- ☐ Reducing the variance of the variables

✔ **Correct**

2. Hierarchical Linear Regression is beneficial when pooling data results in vital group information being lost

1 / 1 point

- ☒ True
- ☐ False

✔ **Correct**

3. Non-hierarchical Linear Regression with groups of sparse data can result in

1 / 1 point

- ☒ very large credible intervals
- ☐ very small credible intervals

✔ **Correct**

4. In Hierarchical Linear Regression, prior parameters for distributions are

1 / 1 point

- ☐ are constant values
- ☒ Distributions

☒ Correct

5. Polynomial Regression is useful for

1 / 1 point

- ☐ Linear data
- ☒ Non-linear data

☒ Correct

6. Multiple Linear Regression is used when you have

1 / 1 point

- ☐ Multiple dependent or target variables
- ☒ Multiple independent or predictor variables

☒ Correct

7. PyMC3 allows you to model multidimensional variables using the shape parameter in a distribution definition, without having to create multiple parameters explicitly.

1 / 1 point

- ☒ True
- ☐ False

☒ Correct

8. The inverse link function of linear regression is

1 / 1 point

- ☐ Logit function
- ☒ Identity function

✓ **Correct**

9. Multiclass classification uses the

1 / 1 point

- ☐ Sigmoid function
- ☒ Softmax function

✓ **Correct**

10. A binary classification problem uses the Bernoulli distribution to model the target, the multiclass classification uses

1 / 1 point

- ☒ Categorical distribution
- ☐ Poisson distribution

✓ **Correct**

11. Does this code indicate a

0 / 1 point

```
 $\alpha_{\text{tmp}} = \text{pm.Normal}(\text{'}\alpha_{\text{tmp}}\text{'}, \text{mu}=2, \text{sd}=5, \text{shape}=M)$   
 $\beta = \text{pm.Normal}(\text{'}\beta\text{'}, \text{mu}=0, \text{sd}=10, \text{shape}=M)$   
 $\epsilon = \text{pm.HalfCauchy}(\text{'}\epsilon\text{'}, 5)$   
 $v = \text{pm.Exponential}(\text{'}v\text{'}, 1/30)$   
 $y_{\text{pred}} = \text{pm.StudentT}(\text{'}y_{\text{pred}}\text{'}, \text{mu}=\alpha_{\text{tmp}}[\text{idx}] + \beta[\text{idx}] * x_{\text{centered}},$   
                     $\text{sd}=\epsilon, \text{nu}=v, \text{observed}=y_{\text{m}})$ 
```

- ☒ Hierarchical model
- ☐ Non-hierarchical model

✗ **Incorrect**

The prior distribution has constant parameters.