

## 6: Generalized Linear Models

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$ echo "Data Science Institute"
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**What happens if we are faced with a situation where the response  $Y$  is neither quantitative or qualitative?**

# Motivation

We have learned about linear and logistic regression which are generalized linear models (GLM). But exactly is GLM? Let's explore!

# What is Generalized Linear Model?

According to Generalized Linear Modelling (GLM) is a flexible generalization of ordinary linear regression. Neither linear regression nor the classification approaches considered so far are applicable.

# Examples of GLM

Here are three examples of GLM:

- Linear Regression
- Logistic Regression
- Poisson Regression

# Poisson Regression

$$\log(\lambda(X_1, \dots, X_p)) = \beta_0 + \beta_1 X_1 + \dots + \beta_p X_p$$

Note: Taking the log ensures that  $\lambda$  can only be non-negative.

This is equivalent to representing the mean  $\lambda$  as follows:

$$\lambda = \mathbb{E}(Y) = \lambda(X_1, \dots, X_p) = e^{\beta_0 + \beta_1 X_1 + \dots + \beta_p X_p}$$

# Exercise: Linear and Poisson Regression on Bikeshare data

## Breakout Room

What are some advantages of Poisson Regression over Linear Regression?



# Common Characteristics of GLM

- Use a set of predictors  $X_1, \dots, X_p$  to predict a response  $Y$
- Model the response  $Y$  as coming from a particular distribution

# References

Chapter 4 of the ISLP book:

James, Gareth, et al. "Classification." An Introduction to Statistical Learning: with Applications in Python, Springer, 2023.