

Ordinal, Generalized ordinal, Poisson Regression Models

```
# Install Library
library(readr)

# Ordinal regression packages
library(VGAM)

# Install and load the "MASS" package if not already installed
if (!require("MASS")) {
  install.packages("MASS")
  library("MASS")
}
```

Rows: 109 Columns: 38

-- Column specification -----
Delimiter: ","

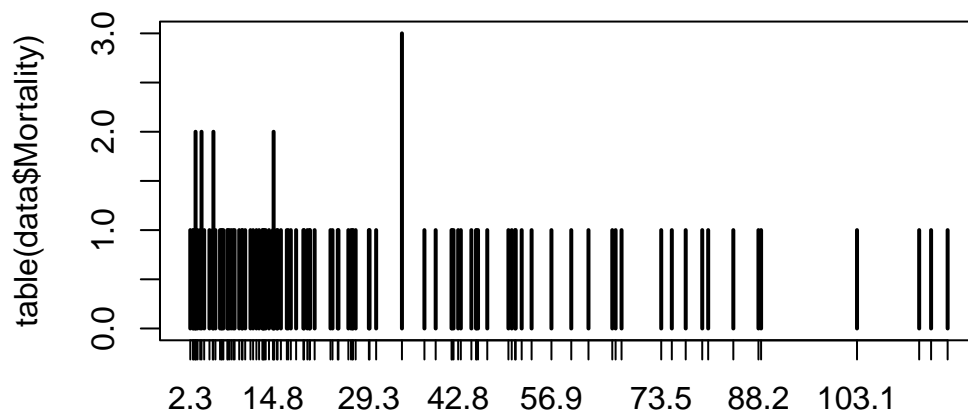
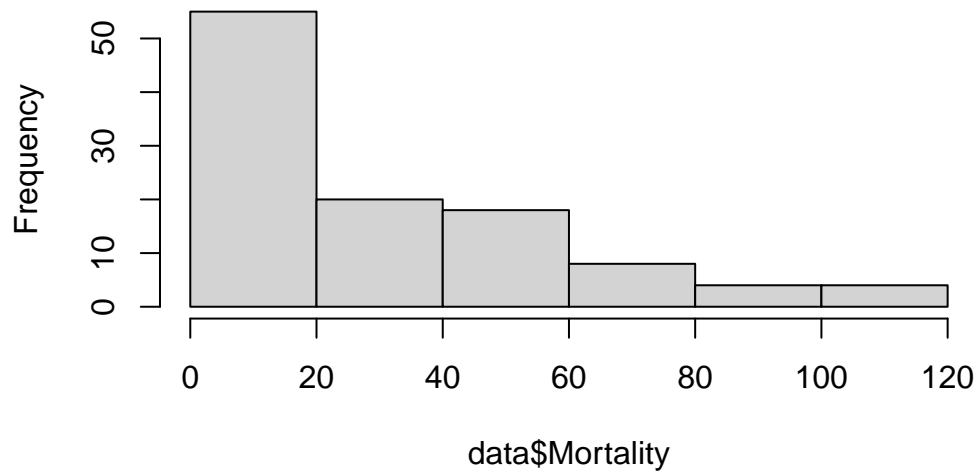
chr (7): Country, CC, pm_category, pm_ordinal, Region, Mortality_Category, ...
dbl (31): CF, GDP, RE, CHE, CHE_GDP, CO2, CO2m, IMD, IMH, IMM, F, Sanitation...

i Use `spec()` to retrieve the full column specification for this data.

i Specify the column types or set `show_col_types = FALSE` to quiet this message.

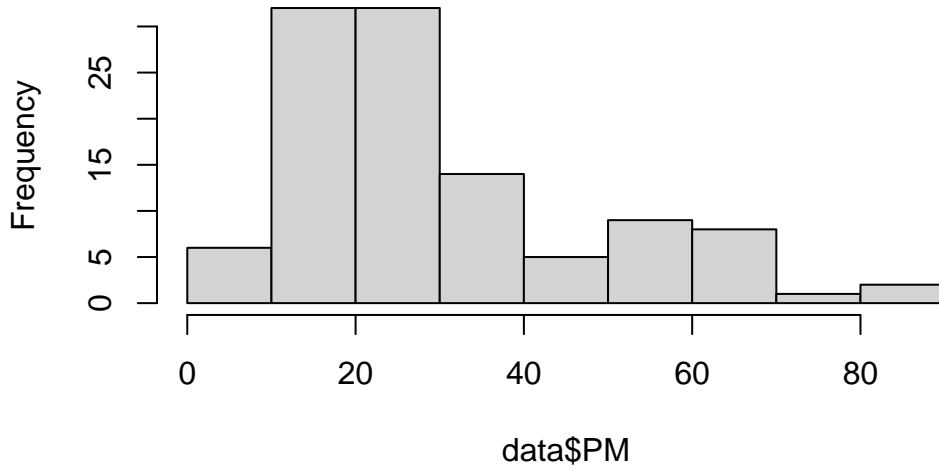
Data Summaries of potential final model variables

Histogram of data\$Mortality

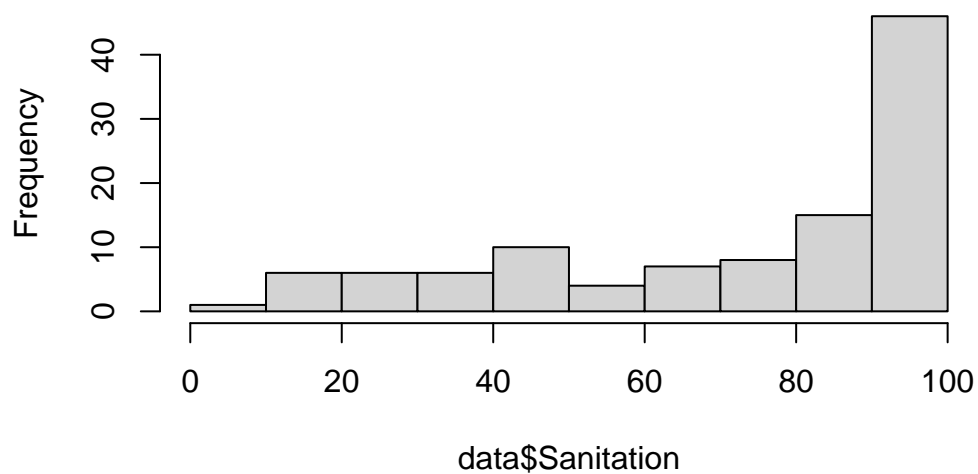


2.3	2.7	2.8	3	3.1	3.2	3.4	3.8	4	4.4	5.2	5.7	5.8
1	1	1	1	2	1	1	1	2	1	1	1	2
6.1	6.8	7	7.1	7.3	7.9	8	8.2	8.5	8.8	9	9.7	10.1
1	1	1	1	1	1	1	1	1	1	1	1	1
10.2	10.6	11.4	11.8	12.3	12.7	13.2	13.3	13.4	13.5	13.7	14.2	14.8
1	1	1	1	1	1	1	1	1	1	1	1	1
14.9	15	15.4	15.5	16	16.9	17.1	17.5	18.3	19.4	19.5	20	20.3
2	1	1	1	1	1	1	1	1	1	1	1	1
20.4	21.1	23.5	23.8	24.6	24.7	26.2	26.6	26.8	26.9	27.3	29.3	29.4
1	1	1	1	1	1	1	1	1	1	1	1	1
30.4	34.3	37.7	39.4	41.8	41.9	42.1	42.8	43.2	44.8	45.5	45.6	45.8
1	3	1	1	1	1	1	1	1	1	1	1	1
47.2	50.4	50.9	51.4	51.5	52.4	53.9	56.9	59.9	62.5	66.1	66.6	67.5
1	1	1	1	1	1	1	1	1	1	1	1	1
73.5	75.1	77.2	79.7	80.6	84.4	88.2	88.6	103.1	112.5	114.3	116.8	
1	1	1	1	1	1	1	1	1	1	1	1	

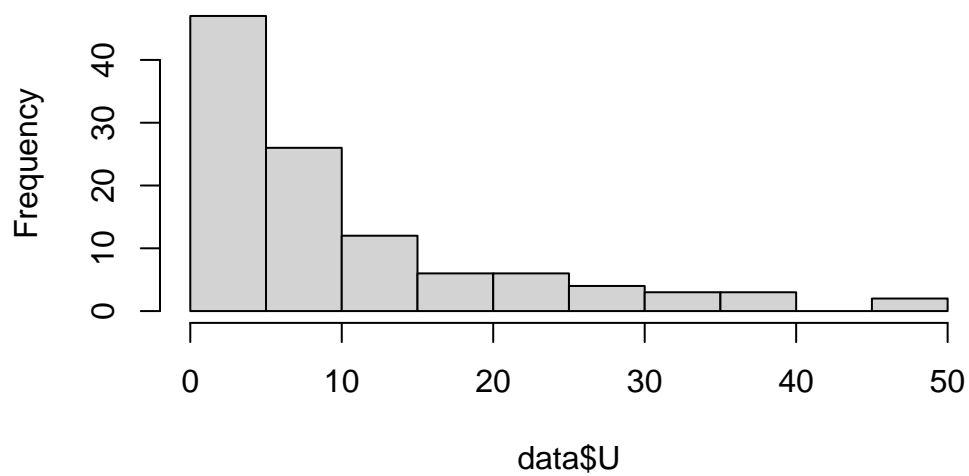
Histogram of data\$PM



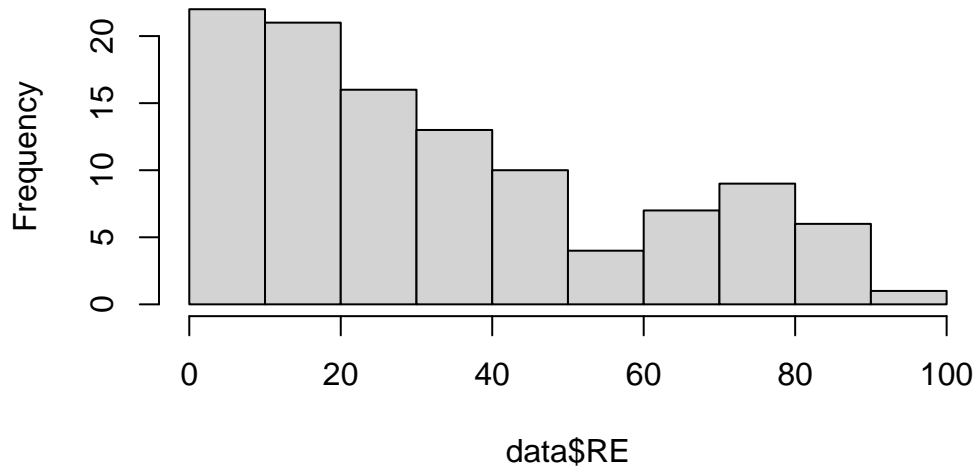
Histogram of data\$Sanitation



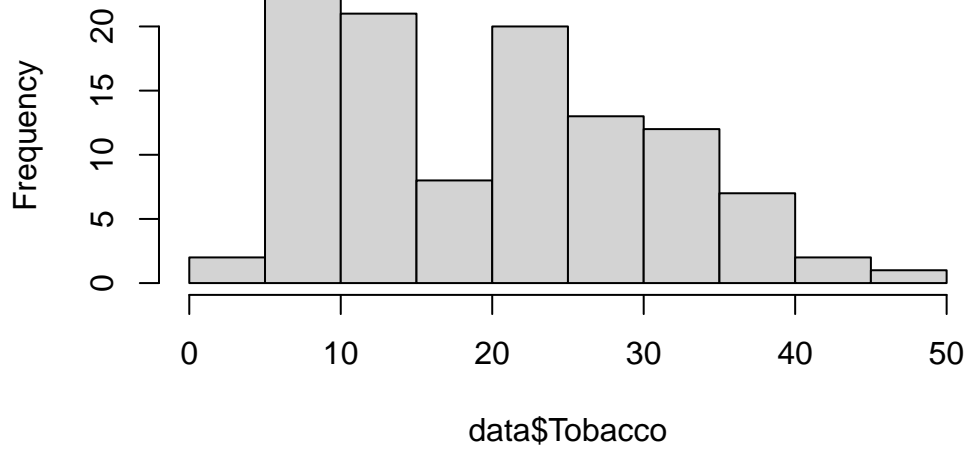
Histogram of data\$U

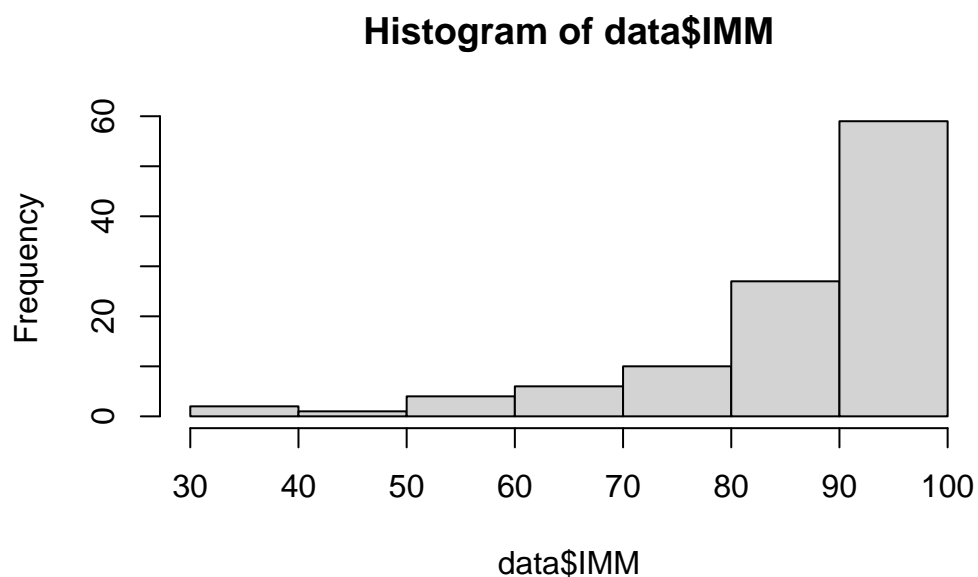


Histogram of data\$RE



Histogram of data\$Tobacco





(5b) cont'd

Ordinal

The ordinal model is not a good fit for the data, as demonstrated below.

Ordinal data formulation:

Refer to the section titled: *Data Formulation To fit a Logistic, Multinomial, Ordinal, Generalized Ordinal, Poisson*

Call:

```
vglm(formula = Mortality_Ordinal ~ PM, family = cumulative(parallel = TRUE,
  reverse = FALSE), data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	-2.03076	0.46224	-4.393	1.12e-05	***
PM	0.05921	0.01436	4.124	3.72e-05	***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Name of linear predictor: logitlink(P[Y<=1])

Residual deviance: 126.2788 on 107 degrees of freedom

Log-likelihood: -63.1394 on 107 degrees of freedom

Number of Fisher scoring iterations: 4

No Hauck-Donner effect found in any of the estimates

Exponentiated coefficients:

PM

1.061003

Call:

```
vglm(formula = Mortality_Ordinal ~ PM + PM_squared, family = cumulative(parallel = TRUE,
reverse = FALSE), data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-2.1703822	0.9596585	-2.262	0.0237 *
PM	0.0689166	0.0598978	1.151	0.2499
PM_squared	-0.0001317	0.0007853	-0.168	0.8668

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Name of linear predictor: logitlink(P[Y<=1])

Residual deviance: 126.2512 on 106 degrees of freedom

Log-likelihood: -63.1256 on 106 degrees of freedom

Number of Fisher scoring iterations: 5

No Hauck-Donner effect found in any of the estimates

Exponentiated coefficients:

	PM	PM_squared
	1.0713468	0.9998683

Call:

```
vglm(formula = Mortality_Ordinal ~ PM + CO2m + PM * CO2m, family = cumulative(parallel = TRUE,
reverse = FALSE), data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-0.4862332	0.8303266	-0.586	0.5581
PM	0.0523685	0.0250454	2.091	0.0365 *
CO2m	-0.5615484	0.3570660	-1.573	0.1158
PM:CO2m	0.0009029	0.0113598	0.079	0.9367

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Name of linear predictor: logitlink(P[Y<=1])

Residual deviance: 98.8368 on 105 degrees of freedom

Log-likelihood: -49.4184 on 105 degrees of freedom

Number of Fisher scoring iterations: 6

No Hauck-Donner effect found in any of the estimates

Exponentiated coefficients:

	PM	CO2m	PM:CO2m
	1.0537639	0.5703253	1.0009033

Call:

```
vglm(formula = Mortality_Ordinal ~ PM + CO2m + GDP + GDP * CO2m,
family = cumulative(parallel = TRUE, reverse = FALSE), data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	1.618e+00	9.204e-01	1.758	0.0788 .
PM	2.354e-02	1.866e-02	1.261	0.2072

CO2m	-2.525e-01	1.864e-01	-1.355	0.1755
GDP	-5.228e-04	1.733e-04	NA	NA
CO2m:GDP	2.458e-05	1.058e-05	2.323	0.0202 *

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Name of linear predictor: logitlink(P[Y<=1])

Residual deviance: 83.4882 on 104 degrees of freedom

Log-likelihood: -41.7441 on 104 degrees of freedom

Number of Fisher scoring iterations: 7

Warning: Hauck-Donner effect detected in the following estimate(s):
'GDP'

Exponentiated coefficients:

	PM	CO2m	GDP	CO2m:GDP
	1.0238153	0.7768465	0.9994773	1.0000246

Call:

```
vglm(formula = Mortality_Ordinal ~ PM + Sanitation + U + +IMM,
     family = cumulative(parallel = TRUE, reverse = FALSE), data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	5.37484	2.82730	1.901	0.05730 .
PM	0.03668	0.01848	1.985	0.04713 *
Sanitation	-0.06028	0.01882	-3.203	0.00136 **
U	0.07496	0.05117	1.465	0.14296
IMM	-0.03325	0.02743	-1.212	0.22552

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Name of linear predictor: logitlink(P[Y<=1])

Residual deviance: 70.4662 on 104 degrees of freedom

Log-likelihood: -35.2331 on 104 degrees of freedom

Number of Fisher scoring iterations: 6

No Hauck-Donner effect found in any of the estimates

Exponentiated coefficients:

	PM	Sanitation	U	IMM
	1.0373585	0.9415007	1.0778434	0.9672985

Call:

```
vglm(formula = Mortality_Ordinal ~ PM + Tobacco + Sanitation +  
      IMM + RE + CO2 + GDP, family = cumulative(parallel = TRUE,  
      reverse = FALSE), data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	5.489e+00	3.238e+00	1.695	0.0900 .
PM	3.233e-02	2.233e-02	1.448	0.1477
Tobacco	-1.110e-02	3.185e-02	-0.349	0.7273
Sanitation	-4.636e-02	2.064e-02	-2.247	0.0247 *
IMM	-2.860e-02	2.806e-02	-1.019	0.3080
RE	1.305e-02	1.717e-02	0.760	0.4472
CO2	-1.917e-07	5.977e-07	-0.321	0.7484
GDP	-1.737e-04	1.250e-04	NA	NA

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Name of linear predictor: logitlink(P[Y<=1])

Residual deviance: 67.6773 on 101 degrees of freedom

Log-likelihood: -33.8386 on 101 degrees of freedom

Number of Fisher scoring iterations: 7

Warning: Hauck-Donner effect detected in the following estimate(s):
'GDP'

Exponentiated coefficients:

PM	Tobacco Sanitation	IMM	RE	CO2	GDP
1.0328540	0.9889580	0.9546942	0.9718014	1.0131316	0.9999998
					0.9998263

Generalized ordinal

The generalized ordinal model is not a good fit for the data, as demonstrated below.

Call:

```
vglm(formula = Mortality_Ordinal ~ PM, family = cumulative(parallel = FALSE,
  reverse = T), data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	2.03076	0.46224	4.393	1.12e-05 ***
PM	-0.05921	0.01436	-4.124	3.72e-05 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Name of linear predictor: logitlink(P[Y>=2])

Residual deviance: 126.2788 on 107 degrees of freedom

Log-likelihood: -63.1394 on 107 degrees of freedom

Number of Fisher scoring iterations: 4

No Hauck-Donner effect found in any of the estimates

Exponentiated coefficients:

PM
0.9425043

Call:

```
vglm(formula = Mortality_Ordinal ~ PM + PM_squared, family = cumulative(parallel = FALSE,
  reverse = T), data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	2.1703822	0.9596585	2.262	0.0237 *
PM	-0.0689166	0.0598978	-1.151	0.2499
PM_squared	0.0001317	0.0007853	0.168	0.8668

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Name of linear predictor: logitlink(P[Y>=2])

Residual deviance: 126.2512 on 106 degrees of freedom

Log-likelihood: -63.1256 on 106 degrees of freedom

Number of Fisher scoring iterations: 5

No Hauck-Donner effect found in any of the estimates

Exponentiated coefficients:

	PM	PM_squared
	0.9334045	1.0001317

Call:

```
vglm(formula = Mortality_Ordinal ~ PM + CO2m + PM * CO2m, family = cumulative(
  parallel = FALSE, reverse = T), data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	0.4862332	0.8303266	0.586	0.5581
PM	-0.0523685	0.0250454	-2.091	0.0365 *
CO2m	0.5615484	0.3570660	1.573	0.1158
PM:CO2m	-0.0009029	0.0113598	-0.079	0.9367

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Name of linear predictor: logitlink(P[Y>=2])

Residual deviance: 98.8368 on 105 degrees of freedom

Log-likelihood: -49.4184 on 105 degrees of freedom

Number of Fisher scoring iterations: 6

No Hauck-Donner effect found in any of the estimates

Exponentiated coefficients:

	PM	CO2m	PM:CO2m
	0.9489791	1.7533854	0.9990975

Call:

```
vglm(formula = Mortality_Ordinal ~ PM + CO2m + GDP + GDP * CO2m,  
      family = cumulative(parallel = FALSE, reverse = T), data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-1.618e+00	9.204e-01	-1.758	0.0788 .
PM	-2.354e-02	1.866e-02	-1.261	0.2072
CO2m	2.525e-01	1.864e-01	1.355	0.1755
GDP	5.228e-04	1.733e-04	NA	NA
CO2m:GDP	-2.458e-05	1.058e-05	-2.323	0.0202 *

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Name of linear predictor: logitlink(P[Y>=2])

Residual deviance: 83.4882 on 104 degrees of freedom

Log-likelihood: -41.7441 on 104 degrees of freedom

Number of Fisher scoring iterations: 7

Warning: Hauck-Donner effect detected in the following estimate(s):
'GDP'

Exponentiated coefficients:

	PM	CO2m	GDP	CO2m:GDP
	0.9767387	1.2872556	1.0005229	0.9999754

Call:

```
vglm(formula = Mortality_Ordinal ~ PM + Sanitation + U + +IMM,
      family = cumulative(parallel = FALSE, reverse = T), data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-5.37484	2.82730	-1.901	0.05730 .
PM	-0.03668	0.01848	-1.985	0.04713 *
Sanitation	0.06028	0.01882	3.203	0.00136 **
U	-0.07496	0.05117	-1.465	0.14296
IMM	0.03325	0.02743	1.212	0.22552

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Name of linear predictor: logitlink(P[Y>=2])

Residual deviance: 70.4662 on 104 degrees of freedom

Log-likelihood: -35.2331 on 104 degrees of freedom

Number of Fisher scoring iterations: 6

No Hauck-Donner effect found in any of the estimates

Exponentiated coefficients:

	PM	Sanitation	U	IMM
	0.9639869	1.0621341	0.9277785	1.0338071

Call:

```
vglm(formula = Mortality_Ordinal ~ PM + Tobacco + Sanitation +
      IMM + RE + CO2, family = cumulative(parallel = FALSE, reverse = T),
      data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-5.254e+00	3.165e+00	-1.660	0.096954 .
PM	-4.401e-02	2.063e-02	-2.134	0.032865 *
Tobacco	2.143e-03	3.020e-02	0.071	0.943432
Sanitation	6.459e-02	1.894e-02	3.409	0.000651 ***
IMM	2.716e-02	2.740e-02	0.991	0.321601
RE	-1.686e-02	1.676e-02	-1.006	0.314553

```
C02          2.842e-07  6.169e-07  0.461 0.645036
```

```
---
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Name of linear predictor: logitlink(P[Y>=2])
```

```
Residual deviance: 71.2439 on 102 degrees of freedom
```

```
Log-likelihood: -35.622 on 102 degrees of freedom
```

```
Number of Fisher scoring iterations: 6
```

```
No Hauck-Donner effect found in any of the estimates
```

```
Exponentiated coefficients:
```

PM	Tobacco Sanitation	IMM	RE	C02
0.9569404	1.0021450	1.0667174	1.0275278	0.9832856

```
Warning in pchisq(deviance(mod.gen.ordinal.4) - deviance(mod.gen.ordinal.3), :  
NaNs produced
```

```
[1] NaN
```

(5c)

Poisson

Poisson data formulation:

Refer to the section titled: *Data Formulation To fit a Logistic, Multinomial, Ordinal, Generalized Ordinal, Poisson*

Outline of analysis setup:

1. Data Preparation:

- Load and clean our dataset, ensuring that it contains the variables of interest: PM2.5 air quality, under-5 mortality (Mortality) rate, Sanitation rate, Undernourishment rate, Renewable Energy Access, Immunization (IMM) rate, and Tobacco usage.

2. Exploratory Data Analysis (EDA):

- Start with an exploratory data analysis to understand the distributions and relationships between variables. This includes summary statistics, correlation analysis, and data visualization.

3. Model Setup:

- Define our outcome variable (dependent variable):
 - Under-5 mortality (Mortality)
- Identify potential confounding variables:
 - Sanitation
 - Undernourishment
 - Renewable Energy
 - Immunization (IMM)
 - Tobacco
- Define our predictor (independent variable):
 - PM2.5 air quality (PM)
- Create a data frame containing the variables of interest.

4. Assumptions Check:

- Check the assumptions of the poisson regression model, including mean = variance assumption.

5. Multiple Linear Regression Model:

- Set up a poisson multiple linear regression model.

6. Model Fitting:

- Fit the multiple linear regression model to your data.

7. Interpretation:

- Examine the coefficients, p-values, and confidence intervals for each predictor in the model.
- Focus on the coefficient for PM2.5 to assess its effect on Mortality while controlling for potential confounders.
- If the coefficient for PM2.5 is statistically significant, it suggests that PM2.5 has an effect on Mortality.

8. Control for Confounding:

- If there is evidence of confounding, you can further explore interactions between variables or consider additional control variables.

9. Model Assessment:

- Evaluate the overall goodness of fit for the model.

Having done the data preparation, exploratory data analysis and model setup, we now check for whether the assumptions of the poisson regression holds.

Checking Mean = Variance Assumption

The mean is:

```
[1] 30.62661
```

The Variance is:

```
var(data$Mortality)
```

```
[1] 771.6242
```

The Dispersion parameter (= residual deviance/df) is:

```
# Dispersion parameter (= residual deviance/df)
qpois <- glm(Mortality ~ PM, data=data, family = quasipoisson)
summary(qpois)$dispersion
```

```
[1] 17.69821
```

The Mean = Variance Assumption does not hold. The variance of this (presumably) Poisson outcome variable is much greater than the mean. Our data exhibit overdispersion. Hence, our data is not truly supportive of a Poisson modelling

```
[1] 15.93499
```

```
[1] 5.962113
```

```
[1] 5.894287
```

Call:

```
glm(formula = Mortality ~ PM + Sanitation + U + Tobacco + Tobacco_Squared +
     RE + RE_Squared + IMM + IMM_Squared + CO2, family = poisson(),
     data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	2.006e+00	3.661e-01	5.480	4.25e-08	***
PM	1.331e-02	1.045e-03	12.737	< 2e-16	***
Sanitation	-1.248e-02	1.189e-03	-10.495	< 2e-16	***
U	1.177e-02	2.348e-03	5.013	5.35e-07	***
Tobacco	-1.891e-02	8.766e-03	-2.157	0.031031	*
Tobacco_Squared	3.859e-04	1.944e-04	1.985	0.047168	*
RE	1.011e-02	2.916e-03	3.466	0.000528	***
RE_Squared	-8.424e-05	2.921e-05	-2.884	0.003922	**
IMM	4.990e-02	9.548e-03	5.226	1.74e-07	***
IMM_Squared	-3.609e-04	6.612e-05	-5.459	4.80e-08	***
CO2	-8.646e-08	3.238e-08	-2.670	0.007582	**

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for poisson family taken to be 1)

Null deviance: 2462.55 on 108 degrees of freedom
 Residual deviance: 584.55 on 98 degrees of freedom
 AIC: Inf

Number of Fisher Scoring iterations: 5

```
[1] 5.96477
```

We might consider a negative binomial model as a possible alternative since our data exhibits over dispersion. Below we fit a negative binomial:

Negative Binomial Models

Call:

```
glm.nb(formula = Mortality ~ PM, data = data, init.theta = 1.740385005,  
       link = log)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	2.554498	0.146631	17.42	< 2e-16 ***
PM	0.025683	0.004169	6.16	7.26e-10 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for Negative Binomial(1.7404) family taken to be 1)

Null deviance: 153.98 on 108 degrees of freedom
Residual deviance: 117.31 on 107 degrees of freedom
AIC: 936.9

Number of Fisher Scoring iterations: 1

Theta: 1.740
Std. Err.: 0.238

2 x log-likelihood: -930.900

Call:

```
glm.nb(formula = Mortality ~ PM + PM_squared, data = data, init.theta = 1.816027869,  
       link = log)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	1.9563598	0.2881134	6.790	1.12e-11 ***
PM	0.0669913	0.0175271	3.822	0.000132 ***
PM_squared	-0.0005363	0.0002153	-2.490	0.012758 *

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for Negative Binomial(1.816) family taken to be 1)

Null deviance: 160.09 on 108 degrees of freedom
Residual deviance: 117.03 on 106 degrees of freedom
AIC: 934.11

Number of Fisher Scoring iterations: 1

Theta: 1.816
Std. Err.: 0.250

2 x log-likelihood: -926.107

Call:

```
glm.nb(formula = Mortality ~ PM + PM_squared + CO2m + PM * CO2m +  
      PM_squared * CO2m, data = data, init.theta = 2.74925526,  
      link = log)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	2.269e+00	3.364e-01	6.746	1.52e-11	***
PM	7.158e-02	1.960e-02	3.652	0.00026	***
PM_squared	-6.544e-04	2.352e-04	-2.782	0.00540	**
CO2m	-6.986e-02	7.029e-02	-0.994	0.32029	
PM:CO2m	-4.691e-03	4.613e-03	-1.017	0.30923	
PM_squared:CO2m	6.855e-05	6.202e-05	1.105	0.26901	

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for Negative Binomial(2.7493) family taken to be 1)

Null deviance: 232.07 on 108 degrees of freedom
Residual deviance: 116.13 on 103 degrees of freedom
AIC: 897.08

Number of Fisher Scoring iterations: 1

Theta: 2.749
Std. Err.: 0.414

Warning while fitting theta: alternation limit reached

2 x log-likelihood: -883.075

Call:

```
vglm(formula = Mortality_Ordinal ~ PM + PM_squared + GDP + CO2m +  
      GDP * CO2m, family = cumulative(parallel = FALSE, reverse = T),  
      data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-2.249e+00	1.540e+00	-1.461	0.144
PM	1.865e-02	8.438e-02	0.221	0.825
PM_squared	-5.437e-04	1.078e-03	-0.504	0.614
GDP	5.329e-04	1.746e-04	NA	NA
CO2m	2.244e-01	1.926e-01	1.165	0.244
GDP:CO2m	-2.399e-05	1.070e-05	-2.242	0.025 *

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Name of linear predictor: logitlink(P[Y>=2])

Residual deviance: 83.2124 on 103 degrees of freedom

Log-likelihood: -41.6062 on 103 degrees of freedom

Number of Fisher scoring iterations: 7

Warning: Hauck-Donner effect detected in the following estimate(s):
'GDP'

Exponentiated coefficients:

	PM	PM_squared	GDP	CO2m	GDP:CO2m
	1.0188244	0.9994564	1.0005330	1.2515252	0.9999760

Call:

```
vglm(formula = Mortality_Ordinal ~ PM + CO2m + GDP + GDP * CO2m,  
      family = cumulative(parallel = FALSE, reverse = T), data = data)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-1.618e+00	9.204e-01	-1.758	0.0788 .
PM	-2.354e-02	1.866e-02	-1.261	0.2072
CO2m	2.525e-01	1.864e-01	1.355	0.1755
GDP	5.228e-04	1.733e-04	NA	NA
CO2m:GDP	-2.458e-05	1.058e-05	-2.323	0.0202 *

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Name of linear predictor: logitlink(P[Y>=2])

Residual deviance: 83.4882 on 104 degrees of freedom

Log-likelihood: -41.7441 on 104 degrees of freedom

Number of Fisher scoring iterations: 7

Warning: Hauck-Donner effect detected in the following estimate(s):
'GDP'

Exponentiated coefficients:

	PM	CO2m	GDP	CO2m:GDP
	0.9767387	1.2872556	1.0005229	0.9999754

Call:

```
glm.nb(formula = Mortality ~ PM + PM_squared + Sanitation + U +  
      RE + RE_Squared + IMM + IMM_Squared, data = data, init.theta = 5.150795942,  
      link = log)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	1.303e+00	1.073e+00	1.215	0.22428
PM	2.363e-02	1.260e-02	1.875	0.06080 .
PM_squared	-1.496e-04	1.503e-04	-0.995	0.31963
Sanitation	-1.734e-02	3.356e-03	-5.168	2.36e-07 ***
U	1.032e-02	6.946e-03	1.485	0.13747
RE	1.614e-03	7.327e-03	0.220	0.82567
RE_Squared	-8.930e-07	8.100e-05	-0.011	0.99120
IMM	7.070e-02	2.734e-02	2.586	0.00972 **

IMM_Squared -4.831e-04 1.860e-04 -2.597 0.00940 **

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for Negative Binomial(5.1508) family taken to be 1)

Null deviance: 393.75 on 108 degrees of freedom
Residual deviance: 115.52 on 100 degrees of freedom
AIC: 845.07

Number of Fisher Scoring iterations: 1

Theta: 5.151
Std. Err.: 0.903

2 x log-likelihood: -825.074

Call:

```
glm.nb(formula = Mortality ~ PM + Sanitation + U + RE + RE_Squared +  
      IMM + IMM_Squared + CO2, data = data, init.theta = 5.229787106,  
      link = log)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	1.482e+00	1.054e+00	1.406	0.15964
PM	1.230e-02	2.907e-03	4.230	2.33e-05 ***
Sanitation	-1.764e-02	3.267e-03	-5.399	6.69e-08 ***
U	1.119e-02	6.863e-03	1.631	0.10297
RE	-1.716e-04	6.980e-03	-0.025	0.98039
RE_Squared	1.088e-05	7.883e-05	0.138	0.89021
IMM	7.112e-02	2.726e-02	2.609	0.00909 **
IMM_Squared	-4.819e-04	1.856e-04	-2.596	0.00943 **
CO2	-7.848e-08	5.288e-08	-1.484	0.13778

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for Negative Binomial(5.2298) family taken to be 1)

Null deviance: 398.58 on 108 degrees of freedom
Residual deviance: 115.58 on 100 degrees of freedom
AIC: 843.84

Number of Fisher Scoring iterations: 1

Theta: 5.230
Std. Err.: 0.921

2 x log-likelihood: -823.844

Call:

```
glm.nb(formula = Mortality ~ PM + Sanitation + U + Tobacco +  
Tobacco_Squared + RE + RE_Squared + IMM + IMM_Squared + GDP,  
data = data, init.theta = 7.917885787, link = log)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	2.801e+00	9.237e-01	3.033	0.00242	**
PM	7.310e-03	2.582e-03	2.831	0.00464	**
Sanitation	-1.158e-02	2.965e-03	-3.904	9.48e-05	***
U	9.504e-03	6.084e-03	1.562	0.11825	
Tobacco	-3.874e-02	1.968e-02	-1.969	0.04896	*
Tobacco_Squared	6.751e-04	4.321e-04	1.562	0.11822	
RE	-4.523e-03	6.178e-03	-0.732	0.46405	
RE_Squared	4.709e-05	6.933e-05	0.679	0.49703	
IMM	5.060e-02	2.346e-02	2.157	0.03102	*
IMM_Squared	-3.513e-04	1.592e-04	-2.206	0.02738	*
GDP	-5.088e-05	8.107e-06	-6.277	3.46e-10	***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for Negative Binomial(7.9179) family taken to be 1)

Null deviance: 548.46 on 108 degrees of freedom
Residual deviance: 109.41 on 98 degrees of freedom
AIC: 808.3

Number of Fisher Scoring iterations: 1

Theta: 7.92
Std. Err.: 1.48

2 x log-likelihood: -784.301


```
glm.nb(formula = Mortality ~ PM + Sanitation + IMM + IMM_Squared +
      GDP, data = data, init.theta = 7.26843058, link = log)
```

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	2.682e+00	8.708e-01	3.080	0.00207	**
PM	7.757e-03	2.510e-03	3.091	0.00200	**
Sanitation	-1.526e-02	2.215e-03	-6.890	5.56e-12	***
IMM	5.097e-02	2.364e-02	2.156	0.03111	*
IMM_Squared	-3.593e-04	1.613e-04	-2.228	0.02589	*
GDP	-4.928e-05	8.040e-06	-6.129	8.82e-10	***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```
Null deviance: 514.64  on 108  degrees of freedom
Residual deviance: 113.20  on 103  degrees of freedom
AIC: 808.68
```

Theta: 7.27
Std. Err.: 1.37

Response: Mortality

	theta	Resid. df	2 x log-lik.	Test	df	LR stat.	Pr(Chi)
1							
2	PM + Sanitation + U + Tobacco + Tobacco_Squared + RE + RE_Squared + IMM + IMM_Squared + GDI						
1	7.268431	103	-794.6807				
2	7.917886	98	-784.3011	1 vs 2	5	10.37962	0.06516624