

Zero inflated Poisson distribution

For Master 2 Math SV

January 2026

1. The data

We study the abundance of fish species at $n = 89$ sites in the Barents Sea (Fosshheim, Nilssen, and Aschan (2006)). The data are available in the file BarentsFish.csv where the first 4 columns correspond to four environmental covariates covariates (latitude, longitude, depth, temperature) and the next 30 columns are the abundances of 30 species.

```
dataCodBarents <- read.table('BarentsFish.csv', sep=';', header=TRUE)
Covariates <- as.matrix(dataCodBarents[, (1:4)])
Counts <- dataCodBarents[, 5:ncol(dataCodBarents)]
j <- 21 # We focus on Species Tr_es
Abundance <- Counts[, j]; Presence <- (Abundance>0)
Cod_Tr_es<- as.data.frame(Covariates)
Cod_Tr_es$Abundance <- Abundance
Cod_Tr_es$Presence <- Presence

j <- 20 # We focus on Species Se_ma
Abundance <- Counts[, j]; Presence <- (Abundance>0)
Cod_Se_ma<- as.data.frame(Covariates)
Cod_Se_ma$Abundance <- Abundance
Cod_Se_ma$Presence <- Presence
```

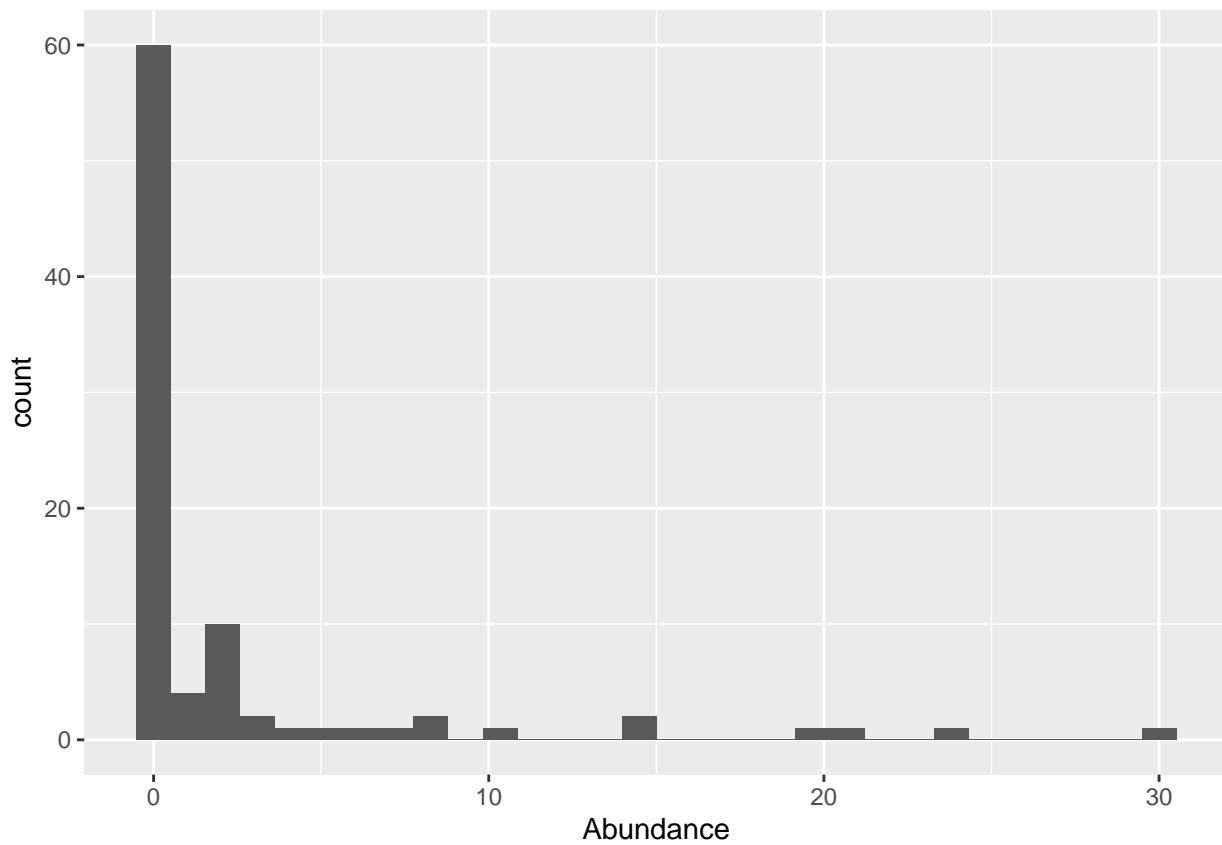
In the following, we will consider only one fish species, for example the 21th ('Tr_es = Trisopterus Esmarkii = Tacaud norvégien) or the 20th ("Se_ma"= Sebastes Marinus) and we will note $1 \leq i \leq n$.

Y_i = abundance of golden redfish in station i.

1. Explore the data with standard tools (means, histograms...) for both species

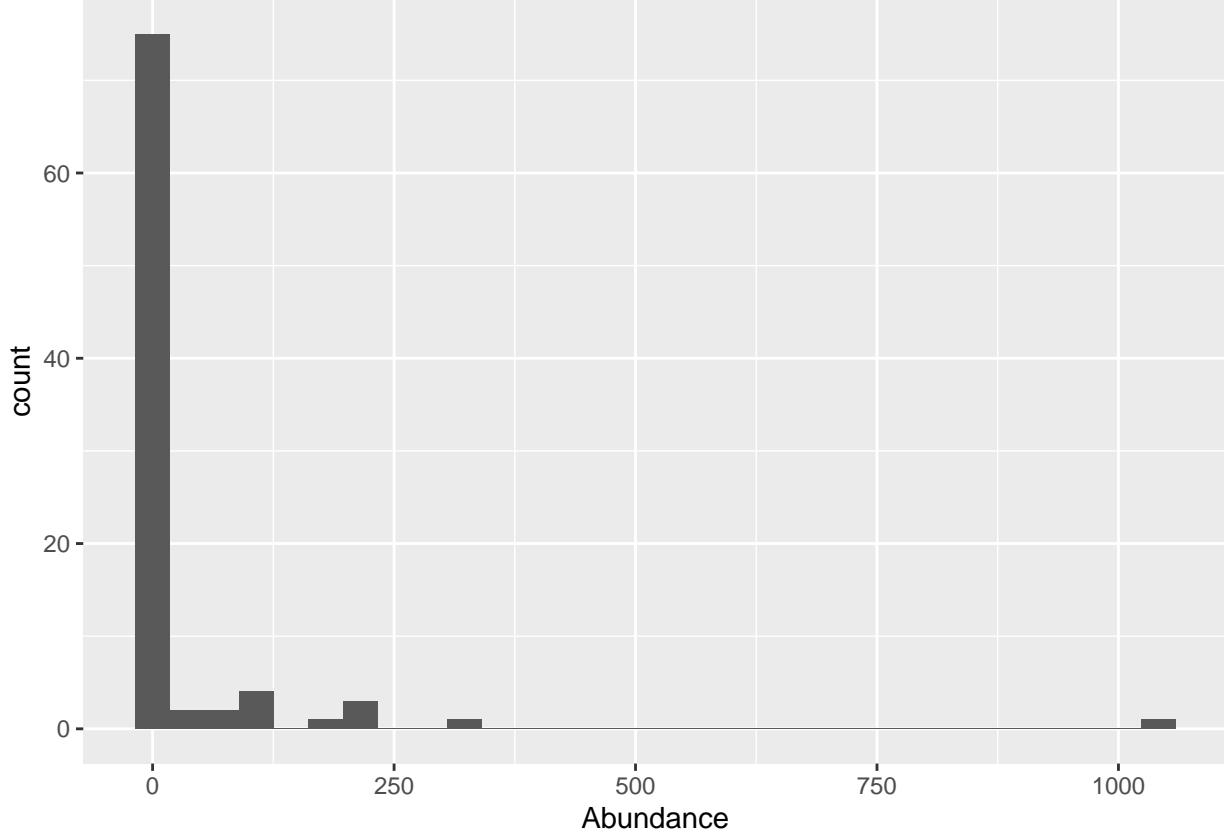
```
library(ggplot2)
ggplot(Cod_Se_ma,aes(x=Abundance))+geom_histogram()

## `stat_bin()` using `bins = 30`. Pick better value `binwidth`.
```



```
ggplot(Cod_Tr_es,aes(x=Abundance))+geom_histogram()
```

```
## `stat_bin()` using `bins = 30`. Pick better value `binwidth`.
```



Observe an over-representation of null values. We propose to modelize this over inflation of 0.

2. Zero-inflated Poisson model

We propose to consider the following Zero Inflation Poisson distribution (ZIP) Let Z_i be a latent variable such that

$$Z_i \sim_{i.i.d} \text{Bern}(\pi)$$

Then

$$Y_i | Z_i \sim (1 - Z_i)\delta_{\{0\}} + Z_i \mathcal{P}(\mu_i) \quad (1)$$

where \mathcal{P} is the Poisson distribution. Z_i represents the presence of the species.

2. Write the marginal distribution of Y_i
3. Derive $\mathbb{E}[Y_i]$ and $P(Y_i = 0)$
4. Write the complete log likelihood $\log p_\theta(\mathbf{Y}, \mathbf{Z})$ of the model where $\theta = (\pi, \mu)$.

We propose to maximize likelihood with respect to the parameters using the EM algorithm

5. Write the corresponding E-step.
6. Write the corresponding M-step.
7. Suggest an initial value for the parameter θ .
8. Code the EM algorithm.
 - a. Test your algorithm on simulated data. Check that the likelihood increases at each iteration of the EM

- b. Test your algorithm on the real data. What happens?

3. ZIP with covariates

We now consider a model similar to ZIP but taking into account the environmental covariates. We note x_i the vector comprising these covariates for the site i :

$$x_i = [1, \text{latitude}_i, \text{longitude}_i, \text{depth}_i, \text{temperature}_i].$$

We therefore pose : $(Z_i)_{1, \leq i \leq n}$ independent, $(Y_i|Z_i)_{1, \leq i \leq n}$ independent and

$$\begin{aligned} Z_i &\sim \text{Bern}(\pi_i) & \text{with } \log\left(\frac{\pi_i}{1-\pi_i}\right) &= x_i^T \alpha \\ Y_i|Z_i &\sim (1 - Z_i)\delta_{\{0\}} + Z_i \mathcal{P}(\mu_i) & \text{with } \log \mu_i &= x_i^T \beta \end{aligned} \quad (2)$$

The vectors α and β contain the regression coefficients to predict absence and abundance conditional on the presence of the species at each site.

9. Write the full log likelihood $p_\theta(\mathbf{Y}, \mathbf{Z})$ of this new model as a function of the parameter $\theta = (\alpha, \beta)$.
10. Write the E-step.
11. Write the M-step. Is it explicit?
12. Propose an initial value for the parameter θ .

References

Fossheim, Maria, Einar M. Nilssen, and Michaela Aschan. 2006. “Fish Assemblages in the Barents Sea.” *Marine Biology Research* 2 (4): 260–69. <https://doi.org/10.1080/17451000600815698>.