

Beamer Presentation – Consommation alimentaire par zone

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```
# Chargement des données depuis le fichier ZIP
```

```
zip_path <- "/Users/HP/Downloads/SEN2018_menage.zip"
```

```
out_dir <- "/Users/HP/Downloads/SEN2018_menage"
```

```
if (!dir.exists(out_dir) && file.exists(zip_path)) unzip(zip_p
```

```
dta_files <- list.files(out_dir, pattern = "\\..dta$", full.nam
```

```
data_list <- setNames(lapply(dta_files, haven::read_dta), base
```

```
# Extraction des variables nécessaires
```

```
roster <- data_list[["s01_me_SEN2018.dta"]] %>% mutate(id = pa
```

```
cons_food <- data_list[["s08b1_me_SEN2018.dta"]] %>% mutate(ic
```

```
food_vars <- grep("^s08b02", names(cons_food), value = TRUE)
```

```
food_totals <- cons_food %>%
```

```
  select(id, all_of(food_vars)) %>%
```

```
  mutate(food_total = rowSums(across(all_of(food_vars)), na.rm
```

```
  select(id, food_total) %>%
```

```
  filter(food_total > 0) %>%
```

```
  distinct(id, .keep_all = TRUE)
```

```
zone_df <- data_list[["s00_me_SEN2018.dta"]] %>%
```

- Étude de la consommation alimentaire individuelle
- Variables : zone (urbaine vs rurale), taille du ménage
- Objectif : détecter des différences et effets significatifs

Statistiques descriptives

```
desc_zone <- df_indiv %>%  
  group_by(zone) %>%  
  summarise(  
    n = n(),  
    moyenne = mean(food_per_indiv),  
    ecart_type = sd(food_per_indiv)  
  )  
knitr::kable(desc_zone, caption = "Statistiques par zone")
```

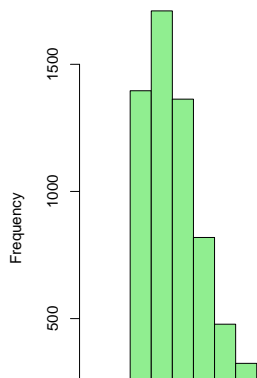
Table 1: Statistiques par zone

zone	n	moyenne	ecart_type
Rurale	3212	7.391587	8.285034
Urbaine	3925	11.384799	13.532003

Histogramme de distribution

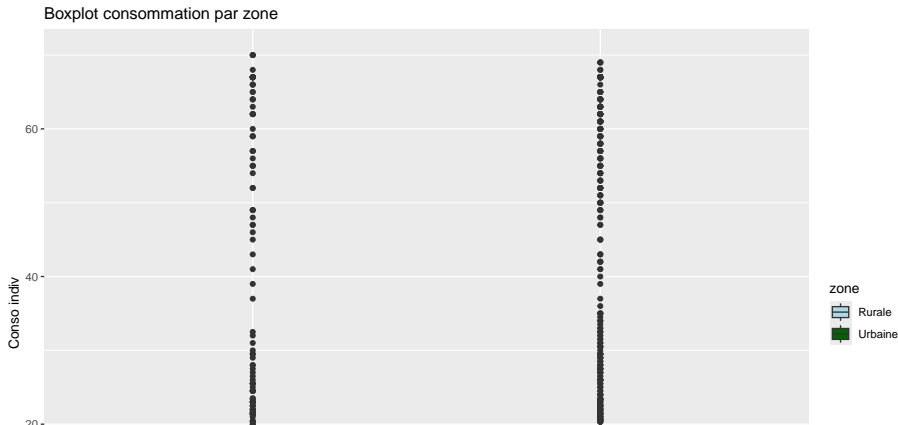
```
hist(df_indiv$food_per_indiv, breaks = 30, col = "lightgreen",  
     main = "Distribution consommation individuelle",  
     xlab = "Consommation par individu")
```

Distribution consommation individuelle



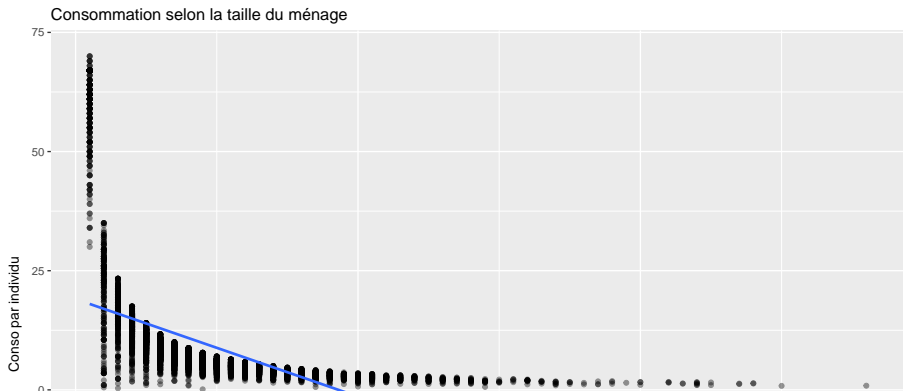
Boxplot par zone

```
ggplot(df_indiv, aes(x = zone, y = food_per_indiv, fill = zone)) +  
  geom_boxplot() +  
  scale_fill_manual(values = c("Urbaine" = "darkgreen", "Rurale" = "darkred")) +  
  labs(title = "Boxplot consommation par zone", x = "Zone", y = "Consommation par individu")
```



Taille ménage vs consommation

```
ggplot(df_indiv, aes(x = n_indiv, y = food_per_indiv)) +  
  geom_point(alpha = 0.4) +  
  geom_smooth(method = "lm", se = FALSE) +  
  labs(title = "Consommation selon la taille du ménage",  
        x = "Nombre d'individus", y = "Conso par individu")
```



Corrélation Spearman

```
cor.test(df_indiv$n_indiv, df_indiv$food_per_indiv, method = 's')
```

```
##  
## Spearman's rank correlation rho  
##  
## data: df_indiv$n_indiv and df_indiv$food_per_indiv  
## S = 1.1538e+11, p-value < 2.2e-16  
## alternative hypothesis: true rho is not equal to 0  
## sample estimates:  
## rho  
## -0.9042487
```

Tests de normalité (Shapiro-Wilk)

```
shapiro.test(df_indiv$food_per_indiv[df_indiv$zone == "Urbaine"])
```

```
##  
##  Shapiro-Wilk normality test  
##  
## data:  df_indiv$food_per_indiv[df_indiv$zone == "Urbaine"]  
## W = 0.56338, p-value < 2.2e-16
```

```
shapiro.test(df_indiv$food_per_indiv[df_indiv$zone == "Rurale"])
```

```
##  
##  Shapiro-Wilk normality test  
##  
## data:  df_indiv$food_per_indiv[df_indiv$zone == "Rurale"]  
## W = 0.49368, p-value < 2.2e-16
```

Homogénéité des variances (Levene)

```
leveneTest(food_per_indiv ~ zone, data = df_indiv)
```

```
## Levene's Test for Homogeneity of Variance (center = median)
```

```
##           Df F value      Pr(>F)
```

```
## group      1 127.42 < 2.2e-16 ***
```

```
##           7135
```

```
## ---
```

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

Test de Wilcoxon

```
wilcox.test(food_per_indiv ~ zone, data = df_indiv)
```

```
##
```

```
## Wilcoxon rank sum test with continuity correction
```

```
##
```

```
## data: food_per_indiv by zone
```

```
## W = 4628035, p-value < 2.2e-16
```

```
## alternative hypothesis: true location shift is not equal to 0
```

Taille d'effet (Cohen's d)

```
cohens_d(df_indiv, food_per_indiv ~ zone)
```

```
## # A tibble: 1 x 7
```

```
##   .y.          group1 group2  effsize    n1    n2 magnitu
```

```
## * <chr>          <chr>  <chr>    <dbl> <int> <int> <ord>
```

```
## 1 food_per_indiv Rurale Urbaine  -0.356  3212  3925 small
```

Régression multivariée

```
model <- lm(log(food_per_indiv + 1) ~ zone + n_indiv, data = c  
summary(model)
```

```
##
```

```
## Call:
```

```
## lm(formula = log(food_per_indiv + 1) ~ zone + n_indiv, data
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max  
## -2.37944 -0.19898 -0.06222  0.10666  2.50878
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t value Pr(>|t|)  
## (Intercept)  2.7684479  0.0109169  253.59  <2e-16 ***  
## zoneUrbaine  0.1477093  0.0097025   15.22  <2e-16 ***  
## n_indiv      -0.0830111  0.0008116 -102.29  <2e-16 ***
```

```
## ---
```

Conclusion

La taille du ménage influence **négativement** la consommation individuelle
La zone **urbaine** montre une consommation plus élevée Le modèle multivarié est significatif ($R^2 > 60\%$) Tests non paramétriques utilisés (données non normales)