

Variable categórica

CEPAL

17/2/2022

Lectura de la base

```
encuesta <- readRDS("../Data/encuesta.rds")
```

Definir diseño de la muestra con srvyr

```
library(srvyr)

diseno <- encuesta %>%
  as_survey_design(
    strata = Stratum,
    ids = PSU,
    weights = wk,
    nest = T
  ) %>% mutate(Uno = 1)
```

Sub-grupos

Extraer sub-grupos de la encuesta.

```
sub_Urbano <- diseno %>% filter(Zone == "Urban")
sub_Rural  <- diseno %>% filter(Zone == "Rural")
sub_Mujer  <- diseno %>% filter(Sex == "Female")
sub_Hombre <- diseno %>% filter(Sex == "Male")
```

Estimación de proporción de urbano y rural

```
(prop_zona <- diseno %>% group_by(Zone) %>%  
  summarise(  
    prop = survey_mean(vartype = c("se","ci"),  
      proportion = TRUE )))
```

| Zone | prop | prop_se | prop_low | prop_upp |
|-------|--------|---------|----------|----------|
| Rural | 0.4798 | 0.0125 | 0.4551 | 0.5047 |
| Urban | 0.5202 | 0.0125 | 0.4953 | 0.5449 |

```
(prop_zona2 <- diseno %>% group_by(Zone) %>%  
  summarise(  
    prop = survey_prop(vartype = c("se","ci") )))
```

| Zone | prop | prop_se | prop_low | prop_upp |
|-------|--------|---------|----------|----------|
| Rural | 0.4798 | 0.0125 | 0.4550 | 0.5047 |
| Urban | 0.5202 | 0.0125 | 0.4953 | 0.5450 |

Proporción de hombres y mujeres en la zona urbana y rural

```
(prop_sexoU <- sub_Urbano %>% group_by(Sex) %>%  
  summarise(  
    n = unweighted(n()),  
    prop = survey_prop(vartype = c("se","ci"))))
```

| Sex | n | prop | prop_se | prop_low | prop_upp |
|--------|-----|--------|---------|----------|----------|
| Female | 718 | 0.5367 | 0.0119 | 0.5128 | 0.5606 |
| Male | 614 | 0.4633 | 0.0119 | 0.4394 | 0.4872 |

```
(prop_sexoR <- sub_Rural %>% group_by(Sex) %>%  
  summarise(  
    n = unweighted(n()),  
    prop = survey_prop(vartype = c("se","ci"))))
```

| Sex | n | prop | prop_se | prop_low | prop_upp |
|--------|-----|--------|---------|----------|----------|
| Female | 633 | 0.5165 | 0.0153 | 0.4858 | 0.5471 |
| Male | 632 | 0.4835 | 0.0153 | 0.4529 | 0.5142 |

Proporción de hombres en la zona urbana y rural

```
(prop_ZonaH <- sub_Hombre %>% group_by(Zone) %>%  
  summarise(  
    n = unweighted(n()),  
    prop = survey_prop(vartype = c("se","ci"))))
```

| Zone | n | prop | prop_se | prop_low | prop_upp |
|-------|-----|--------|---------|----------|----------|
| Rural | 632 | 0.4905 | 0.0149 | 0.461 | 0.520 |
| Urban | 614 | 0.5095 | 0.0149 | 0.480 | 0.539 |

Proporción de hombres en la zona urbana y rural

```
(prop_ZonaM <- sub_Mujer %>% group_by(Zone) %>%  
  summarise(  
    n = unweighted(n()),  
    prop = survey_prop(vartype = c("se", "ci"))))
```

| Zone | n | prop | prop_se | prop_low | prop_upp |
|-------|-----|--------|---------|----------|----------|
| Rural | 633 | 0.4702 | 0.0167 | 0.4372 | 0.5033 |
| Urban | 718 | 0.5298 | 0.0167 | 0.4967 | 0.5628 |

Estimación de la proporción de personas menor a 18 años

```
diseno %>% mutate(edad_18 = case_when(  
  Age<18~"menor a 18 años",  
  TRUE ~ "mayor o igual a 18 años")) %>%  
group_by(edad_18) %>%  
summarise(  
  Prop = survey_prop(vartype = c("se", "ci")) %>%  
  data.frame())
```

| edad_18 | Prop | Prop_se | Prop_low | Prop_upp |
|-------------------------|--------|---------|----------|----------|
| mayor o igual a 18 años | 0.6764 | 0.0118 | 0.6531 | 0.6997 |
| menor a 18 años | 0.3236 | 0.0118 | 0.3003 | 0.3469 |

Estimación de la proporción de personas menor a 18 años en zona rural

```
sub_Rural %>% mutate(edad_18 = case_when(  
  Age<18~"menor a 18 años",  
  TRUE ~ "mayor o igual a 18 años")) %>%  
group_by(edad_18) %>%  
summarise(  
  Prop = survey_prop(vartype = c("se", "ci")) %>%  
  data.frame())
```

| edad_18 | Prop | Prop_se | Prop_low | Prop_upp |
|-------------------------|--------|---------|----------|----------|
| mayor o igual a 18 años | 0.6442 | 0.0201 | 0.6040 | 0.6844 |
| menor a 18 años | 0.3558 | 0.0201 | 0.3156 | 0.3960 |

Estimación de la proporción de mujeres rango de edad

```
sub_Mujer %>% mutate(edad_rango = case_when(  
  Age >= 18 & Age <= 35 ~ "18 - 35",  
  TRUE ~ "Otro")) %>%  
  group_by(edad_rango) %>%  
  summarise(  
    Prop = survey_prop(vartype = c("se", "ci")) %>%  
    data.frame())
```

| edad_rango | Prop | Prop_se | Prop_low | Prop_upp |
|------------|--------|---------|----------|----------|
| 18 - 35 | 0.3059 | 0.0122 | 0.2818 | 0.3299 |
| Otro | 0.6941 | 0.0122 | 0.6701 | 0.7182 |

Tabla Zona Vs Sexo

```
(  
  prop_sexo_zona <- diseno %>%  
    group_by(Sex, Zone) %>%  
    summarise(  
      prop = survey_prop(vartype = c("se", "ci")),  
      n = unweighted(n())) %>%  
    data.frame()  
)
```

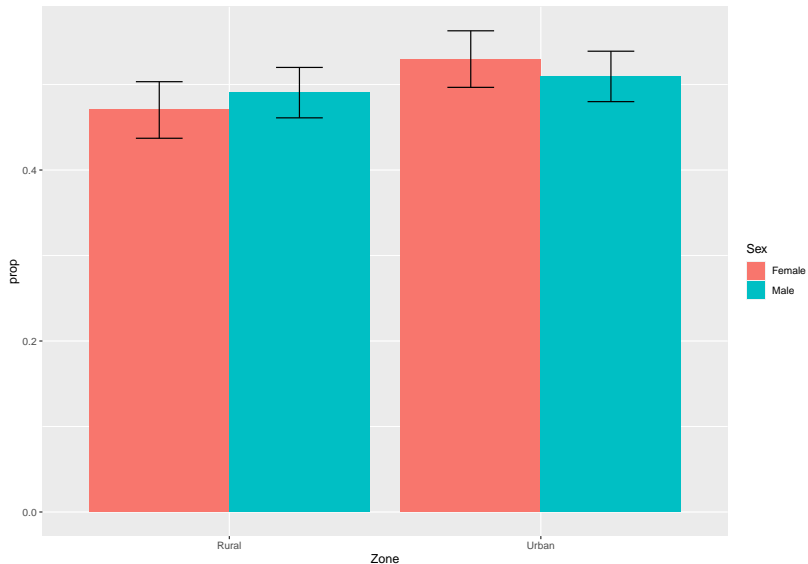
Tabla Zona Vs Sexo

| Sex | Zone | prop | prop_se | prop_low | prop_upp | n |
|--------|-------|--------|---------|----------|----------|-----|
| Female | Rural | 0.4702 | 0.0167 | 0.4372 | 0.5033 | 633 |
| Female | Urban | 0.5298 | 0.0167 | 0.4967 | 0.5628 | 718 |
| Male | Rural | 0.4905 | 0.0149 | 0.4610 | 0.5200 | 632 |
| Male | Urban | 0.5095 | 0.0149 | 0.4800 | 0.5390 | 614 |

Diagrama de barras Zona Vs Sexo

```
require(ggplot2)
ggplot(data = prop_sexo_zona,
      aes(
        x = Zone,
        y = prop,
        ymax = prop_upp,
        ymin = prop_low,
        fill = Sex
      )) +
  geom_bar(stat = "identity", position = "dodge") +
  geom_errorbar(position = position_dodge(width = 0.9),
               width = 0.3)
```

Diagrama de barras Zona Vs Sexo



Prueba de independencia.

```
svychisq(~Sex + Zone, diseno, statistic="F")
```

```
##
```

```
## Pearson's  $X^2$ : Rao & Scott adjustment
```

```
##
```

```
## data: NextMethod()
```

```
## F = 1.1, ndf = 1, ddf = 119, p-value = 0.3
```


Prueba de independencia.

```
svychisq(~Sex + Zone,  
         subset(diseno, Region == "Centro"),  
         statistic="F")
```

```
##
```

```
## Pearson's X^2: Rao & Scott adjustment
```

```
##
```

```
## data:  NextMethod()
```

```
## F = 2.6, ndf = 1, ddf = 23, p-value = 0.1
```

Prueba de independencia.

```
svychisq(~Sex + Zone,  
         subset(diseno, Region == "Sur"),  
         statistic="F")
```

```
##
```

```
## Pearson's X2: Rao & Scott adjustment
```

```
##
```

```
## data:  NextMethod()
```

```
## F = 0.8, ndf = 1, ddf = 24, p-value = 0.4
```

Prueba de independencia.

```
diseno %>% filter(Employment == "Inactive") %>%  
  group_by(Sex) %>%  
  summarise(  
    prop = survey_prop(vartype = c("se", "ci")))
```

| Sex | prop | prop_se | prop_low | prop_upp |
|--------|--------|---------|----------|----------|
| Female | 0.7668 | 0.0135 | 0.7401 | 0.7935 |
| Male | 0.2332 | 0.0135 | 0.2065 | 0.2599 |

Prueba de independencia.

```
svychisq(~Sex + Zone,  
         subset(diseno, Employment == "Inactive"),  
         statistic="F")
```

```
##
```

```
## Pearson's X^2: Rao & Scott adjustment
```

```
##
```

```
## data:  NextMethod()
```

```
## F = 15, ndf = 1, ddf = 105, p-value = 2e-04
```

Tabla con más de dos categorías Sexo Vs empleo.

```
(  
  prop_sexo_empleo <- diseno %>%  
    filter(!is.na(Employment)) %>%  
    group_by(Sex, Employment) %>%  
    summarise(  
      prop = survey_prop(vartype = c("se", "ci")),  
      n = unweighted(n())) %>%  
    data.frame()  
)
```

Tabla del gasto Sexo Vs Rango de edad.

| Sex | Employment | prop | prop_se | prop_low | prop_upp | n |
|--------|------------|--------|---------|----------|----------|-----|
| Female | Unemployed | 0.0285 | 0.0061 | 0.0165 | 0.0406 | 28 |
| Female | Inactive | 0.5505 | 0.0188 | 0.5132 | 0.5878 | 555 |
| Female | Employed | 0.4210 | 0.0188 | 0.3838 | 0.4581 | 458 |
| Male | Unemployed | 0.0724 | 0.0125 | 0.0477 | 0.0971 | 56 |
| Male | Inactive | 0.2037 | 0.0146 | 0.1748 | 0.2326 | 178 |
| Male | Employed | 0.7239 | 0.0207 | 0.6830 | 0.7649 | 645 |

Diagrama de barras del gasto Sexo Vs empleo.

```
require(ggplot2)
ggplot(data = prop_sexo_empleo,
       aes(
         x = Sex,
         y = prop,
         ymax = prop_upp,
         ymin = prop_low,
         fill = Employment
       )) +
  geom_bar(stat = "identity", position = "dodge") +
  geom_errorbar(position = position_dodge(width = 0.9),
               width = 0.3)
```

Diagrama de barras del gasto Sexo Vs Rango de edad.

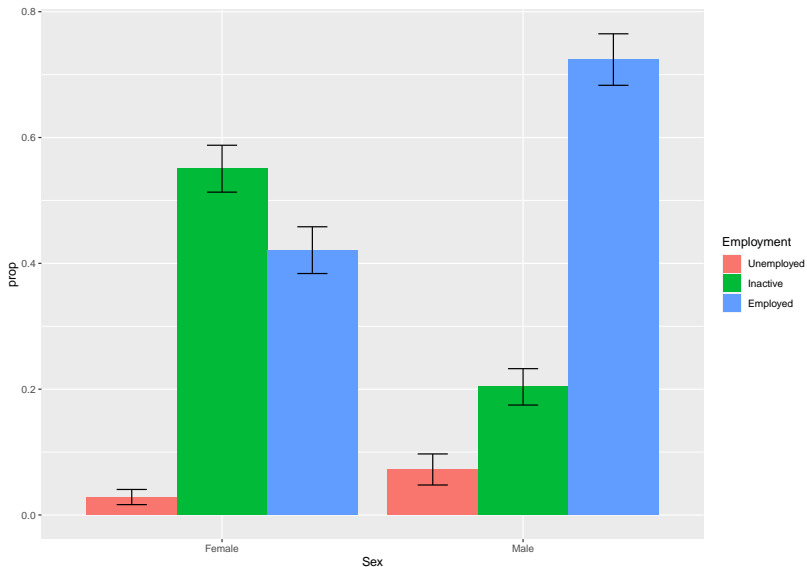


Tabla con más de dos categorías Sexo Vs Region

```
(  
  prop_region_sexo <- diseno %>%  
    group_by(Region, Sex) %>%  
    summarise(  
      prop = survey_prop(vartype = c("se", "ci")),  
      n = unweighted(n())) %>%  
    data.frame()  
)
```

Tabla con más de dos categorías Sexo Vs Region

| Region | Sex | prop | prop_se | prop_low | prop_upp | n |
|-----------|--------|--------|---------|----------|----------|-----|
| Norte | Female | 0.4985 | 0.0161 | 0.4667 | 0.5303 | 285 |
| Norte | Male | 0.5015 | 0.0161 | 0.4697 | 0.5333 | 263 |
| Sur | Female | 0.5390 | 0.0241 | 0.4914 | 0.5867 | 290 |
| Sur | Male | 0.4610 | 0.0241 | 0.4133 | 0.5086 | 261 |
| Centro | Female | 0.5397 | 0.0187 | 0.5027 | 0.5767 | 240 |
| Centro | Male | 0.4603 | 0.0187 | 0.4233 | 0.4973 | 210 |
| Occidente | Female | 0.5161 | 0.0156 | 0.4851 | 0.5470 | 275 |
| Occidente | Male | 0.4839 | 0.0156 | 0.4530 | 0.5149 | 260 |
| Oriente | Female | 0.5381 | 0.0247 | 0.4891 | 0.5870 | 261 |
| Oriente | Male | 0.4619 | 0.0247 | 0.4130 | 0.5109 | 252 |

Diagrama de barras con más de dos categorías Sexo Vs Region

```
ggplot(data = prop_region_sexo,  
      aes(  
        x = Region,  
        y = prop,  
        ymax = prop_upp,  
        ymin = prop_low,  
        fill = Sex  
      )) +  
  geom_bar(stat = "identity", position = "dodge") +  
  geom_errorbar(position = position_dodge(width = 0.9),  
               width = 0.3)
```

Diagrama de barras del gasto Sexo Vs Región

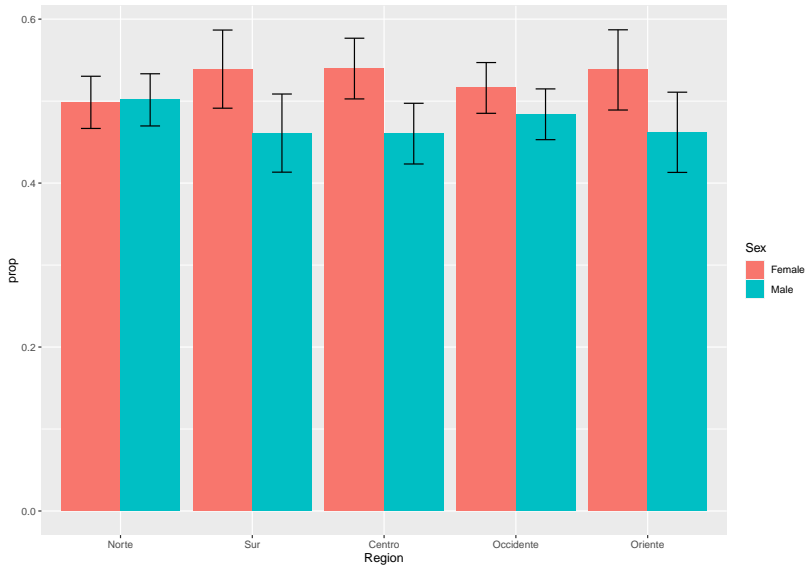


Tabla con más de dos categorías Sexo Vs Region zona urbana

```
(  
  prop_region_sexo_zonaU <- sub_Urbano %>%  
    group_by(Region, Sex) %>%  
    summarise(  
      prop = survey_prop(vartype = c("se", "ci")),  
      n = unweighted(n())) %>%  
    data.frame()  
)
```

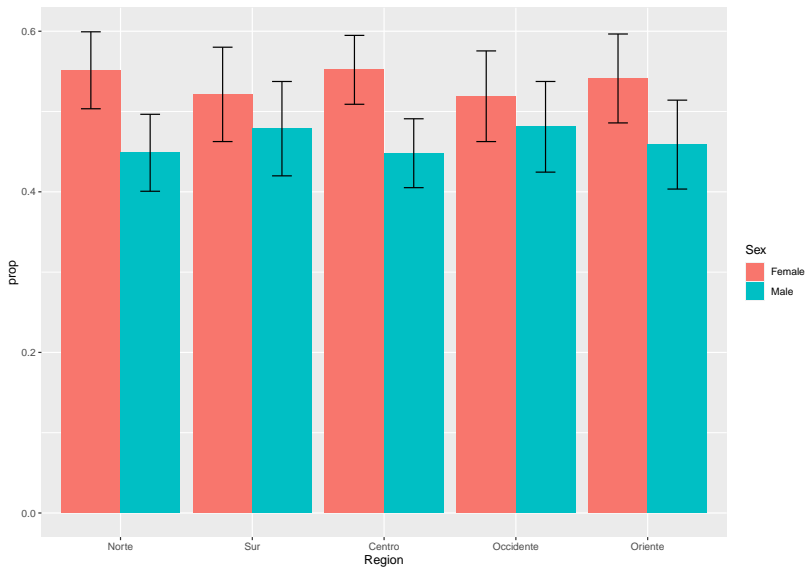
Tabla del gasto Sexo Vs Rango de edad.

| Region | Sex | prop | prop_se | prop_low | prop_upp | n |
|-----------|--------|--------|---------|----------|----------|-----|
| Norte | Female | 0.5514 | 0.0240 | 0.5034 | 0.5993 | 143 |
| Norte | Male | 0.4486 | 0.0240 | 0.4007 | 0.4966 | 111 |
| Sur | Female | 0.5213 | 0.0294 | 0.4626 | 0.5801 | 144 |
| Sur | Male | 0.4787 | 0.0294 | 0.4199 | 0.5374 | 131 |
| Centro | Female | 0.5519 | 0.0215 | 0.5090 | 0.5948 | 167 |
| Centro | Male | 0.4481 | 0.0215 | 0.4052 | 0.4910 | 135 |
| Occidente | Female | 0.5190 | 0.0283 | 0.4626 | 0.5755 | 131 |
| Occidente | Male | 0.4810 | 0.0283 | 0.4245 | 0.5374 | 116 |
| Oriente | Female | 0.5412 | 0.0277 | 0.4858 | 0.5966 | 133 |
| Oriente | Male | 0.4588 | 0.0277 | 0.4034 | 0.5142 | 121 |

Diagrama de barras del gasto Sexo Vs Región

```
ggplot(data = prop_region_sexo_zonaU,  
       aes(  
         x = Region,  
         y = prop,  
         ymax = prop_upp,  
         ymin = prop_low,  
         fill = Sex  
       )) +  
  geom_bar(stat = "identity", position = "dodge") +  
  geom_errorbar(position = position_dodge(width = 0.9),  
               width = 0.3)
```

Diagrama de barras del gasto Sexo Vs Región



glm del Zona en función del ingreso

```
mod <- svyglm(as.factor(Zone) ~ Income,  
              design=disenio, family=quasibinomial)  
summary(mod)$coefficients
```

| | Estimate | Std. Error | t value | Pr(> t) |
|-------------|----------|------------|---------|----------|
| (Intercept) | -0.8300 | 0.1555 | -5.336 | 0 |
| Income | 0.0018 | 0.0003 | 5.564 | 0 |

glm del Zona en función del ingreso

```
mod2 <- svyglm(as.factor(Sex) ~ Income,  
               design=diseño, family=quasibinomial)  
summary(mod2)$coefficients
```

| | Estimate | Std. Error | t value | Pr(> t) |
|-------------|----------|------------|---------|----------|
| (Intercept) | -0.1476 | 0.0736 | -2.0040 | 0.0474 |
| Income | 0.0001 | 0.0001 | 0.7627 | 0.4472 |

Tablas con más de dos categorías.

```
(prop_edad <- diseno %>% group_by(CatAge) %>%  
  summarise(  
    prop = survey_mean(  
      vartype = c("se", "ci"))))
```

| CatAge | prop | prop_se | prop_low | prop_upp |
|-----------|--------|---------|----------|----------|
| 0-5 | 0.0977 | 0.0059 | 0.0860 | 0.1094 |
| 6-15 | 0.1902 | 0.0099 | 0.1705 | 0.2099 |
| 16-30 | 0.2569 | 0.0105 | 0.2361 | 0.2777 |
| 31-45 | 0.2070 | 0.0095 | 0.1881 | 0.2258 |
| 46-60 | 0.1333 | 0.0089 | 0.1156 | 0.1510 |
| Más de 60 | 0.1150 | 0.0089 | 0.0974 | 0.1325 |