# Visualizing co-occurrences of diagnoses in a patient dataset to see patterns by means of Upset Plots

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### **Problem statement**

Contingency tables are in general used to evaluate or cross the simultaneous combination of many characteristics in data. While crossing two qualitative variables might be quite easy to visualize in tables, beyond three variables the task appears to be quite confusing. However in medical analyses, especially for diagnoses, it might be quite helpful to investigate in patients the co-occurrence of medical diagnoses in order to reveal clinical patterns and orientate the decision making. Upset plots turn out to be an ideal tool to visualize and investigate the combination of many diagnoses as well as symptoms in clinical patients. In the framework of this demonstration I will visualize the co-occurrence of diagnoses in patients as well as investigating some of the characteristics of the patients in each combination of diagnoses.

INPUT: an open source patient dataset file in csv format nhefs.csv made available by the University of Havard

A code book of the variables in the dataset is also provided in the file entitled NHEFS\_Codebook.csv

OUTPUT: a PDF file, displaying the UpSet Plot of the combinations of most frequent diagnoses, annotated by other characteristics to show patterns

## **Dataset description**

For the purpose of this project, I will use the nhefs dataset which is a cleaned data set of the data used in Causal Inference by Hernán and Robins. nhefs is dataset containing data from the National Health and Nutrition Examination Survey Data I Epidemiologic Follow-up Study (NHEFS). The NHEFS was jointly initiated by the National Center for Health Statistics and the National Institute on Aging in collaboration with other agencies of the United States Public Health Service. A detailed description of the NHEFS, together with publicly available data sets and documentation, can be found at <a href="https://www.hsph.harvard.edu/miguel-hernan/causal-inference-book/">https://www.hsph.harvard.edu/miguel-hernan/causal-inference-book/</a>

## **Loading libraries and dataset**

options(stringsAsFactors=FALSE)

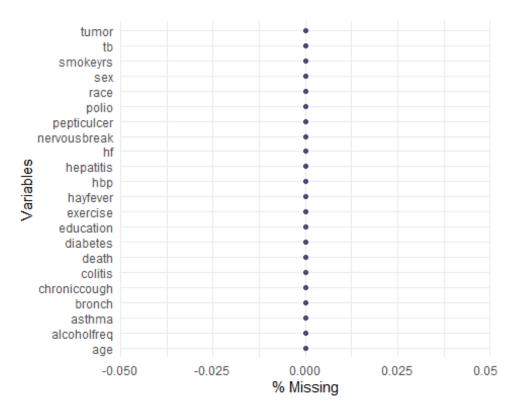
```
if(!require("ComplexUpset"))
    install_github("https://github.com/cran/ComplexUpset")
if(!require("naniar"))
{
    install_github("https://github.com/cran/naniar")
if(!require("ggplot2"))
{
    install_github("https://github.com/cran/ggplot2âPP")
}
if(!require("patchwork"))
{
    install_github("https://github.com/thomasp85/patchwork")
library(devtools)
## Loading required package: usethis
library(ComplexUpset) # to make the Upset Plot
library(ggplot2)
                       # to decorate the Upset Plot
library(patchwork)
                       # to produce the final plot
library(lubridate)
                     # for date functions
library(dplyr)
library(naniar)
file="nhefs.csv"
data=read.csv(file, colClasses="character")
# Displaying the names of the variables in the dataset
colnames(data)
##
   [1] "seqn"
                             "qsmk"
                                                 "death"
##
   [4] "yrdth"
                             "modth"
                                                 "dadth"
   [7] "sbp"
                             "dbp"
                                                 "sex"
## [10] "age"
                             "race"
                                                 "income"
## [13] "marital"
                            "school"
                                                 "education"
## [16] "ht"
                             "wt71"
                                                 "wt82"
## [19] "wt82 71"
                             "birthplace"
                                                 "smokeintensity"
## [22] "smkintensity82 71" "smokeyrs"
                                                 "asthma"
                                                 "hf"
## [25] "bronch"
                             "tb"
## [28] "hbp"
                             "pepticulcer"
                                                 "colitis"
## [31] "hepatitis"
                             "chroniccough"
                                                 "hayfever"
## [34] "diabetes"
                             "polio"
                                                 "tumor"
```

```
## [37] "nervousbreak"
                              "alcoholpy"
                                                   "alcoholfreq"
## [40] "alcoholtype"
                              "alcoholhowmuch"
                                                   "pica"
## [43] "headache"
                              "otherpain"
                                                   "weakheart"
## [46] "allergies"
                              "nerves"
                                                   "lackpep"
## [49] "hbpmed"
                              "boweltrouble"
                                                   "wtloss"
                              "active"
## [52] "infection"
                                                   "exercise"
                              "pregnancies"
                                                   "cholesterol"
## [55] "birthcontrol"
## [58] "hightax82"
                              "price71"
                                                   "price82"
## [61] "tax71"
                              "tax82"
                                                   "price71 82"
## [64] "tax71 82"
```

### Data exploration and visualization of some characteristics

head(data, n=2) seqn qsmk death yrdth modth dadth sbp dbp sex age race income marital school ## 1 233 0 0 175 96 42 19 2 7 2 ## 2 235 0 123 80 36 18 9 ## education ht wt71 wt82 wt82 71 birthplace smokeintensity 1 174.1875 79.04 68.94604024 -10.09395976 47 30 ## 2 2 159.375 58.63 61.23496995 2.60496995 42 20 smkintensity82 71 smokeyrs asthma bronch tb hf hbp pepticulcer colitis ## ## 1 -10 29 0 0 0 0 1 1 ## 2 -10 24 0 0 0 0 hepatitis chroniccough hayfever diabetes polio tumor nervousbreak alcoholpy 0 ## 1 0 1 ## 2 0 0 0 0 1 alcoholfreq alcoholtype alcoholhowmuch pica headache otherpain weakheart ## 1 1 3 7 0 1 0 ## 2 0 1 4 0 1 0 0 allergies nerves lackpep hbpmed boweltrouble wtloss infection active exercise ## 1 0 1 0 ## 2 0 0 0 0 0 1 0 birthcontrol pregnancies cholesterol hightax82 price82 ## price71 ## 1 2 197 0 2.18359375 1.7399902344 ## 2 2 301 0 2.3466796875 1.7973632813 price71 82 tax71 tax82 tax71 82 ## 1 1.1022949219 0.4619750977 0.4437866211 0.6403808594 ## 2 1.3649902344 0.5718994141 0.5493164063 0.79296875

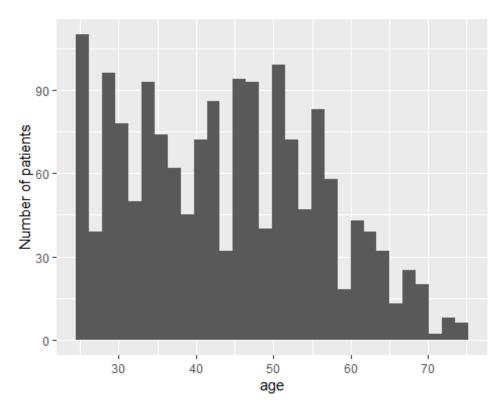
```
diseases <- c( "asthma", "bronch", "tb", "hf", "hbp",</pre>
"pepticulcer", "colitis", "hepatitis", "chroniccough", "hayfever", "diabetes",
"polio", "tumor", "nervousbreak")
# since we are just interested in diseases, I subselect the corresponding
variables with other additional features in a new data frame
subdata <- data[, c(diseases, "death", "smokeyrs", "sex", "age",</pre>
"alcoholfreq", "education", "race", "exercise")]
# investigating the presence of missing data
colSums(is.na (subdata))
##
         asthma
                       bronch
                                                       hf
                                         tb
                                                                   hbp
pepticulcer
##
               0
                            0
                                          0
                                                        0
                                                                      0
0
##
        colitis
                    hepatitis chroniccough
                                                hayfever
                                                              diabetes
polio
               0
                                                        0
##
                                                                      0
##
          tumor nervousbreak
                                      death
                                                smokeyrs
                                                                    sex
age
##
              0
                            0
                                          0
                                                        0
                                                                      0
0
    alcoholfreq
##
                    education
                                       race
                                                exercise
##
                                          0
## Forcing conversion of some numerical variables loaded as strings
subdata[c(diseases, "smokeyrs", "age")] <-</pre>
lapply(subdata[c(diseases, "smokeyrs", "age")], as.numeric)
gg_miss_var(subdata, show_pct = TRUE)
```



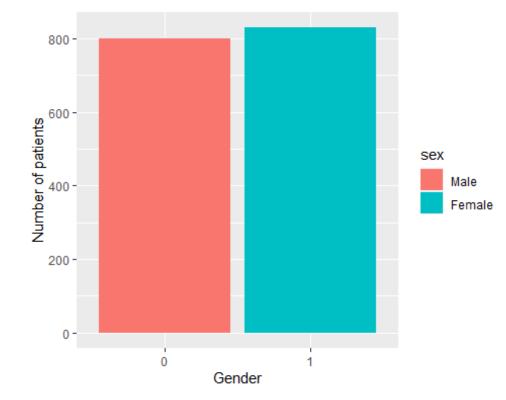
```
par (mfrow=(c(2,2)))

## Age distribution in the dataset

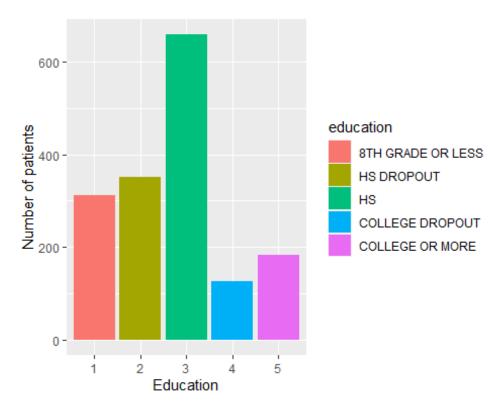
ggplot(subdata, aes(x=age)) +
   geom_histogram()+
   ylab("Number of patients")
```



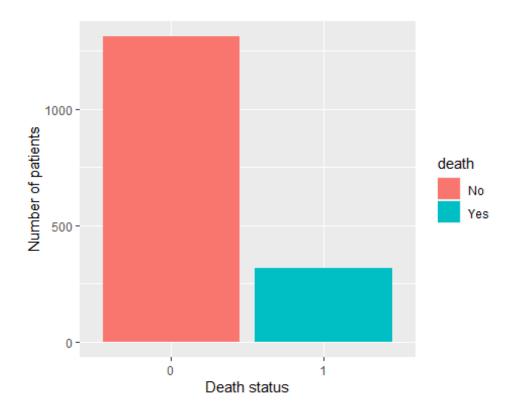
```
ggplot(subdata, aes(x=sex, fill=sex)) +
  geom_bar() +
  scale_fill_discrete(labels=c("Male", "Female"))+
  xlab("Gender")+
  ylab("Number of patients")
```



# 1: 8TH GRADE OR LESS, 2: HS DROPOUT, 3: HS, 4:COLLEGE DROPOUT, 5: COLLEGE
OR MORE
ggplot(subdata, aes(x=education, fill=education)) +
 geom\_bar() +
 scale\_fill\_discrete(labels=c("8TH GRADE OR LESS", "HS DROPOUT", "HS",
"COLLEGE DROPOUT", "COLLEGE OR MORE"))+
 xlab("Education")+
 ylab("Number of patients")



```
ggplot(subdata, aes(x=death, fill=death)) +
  geom_bar() +
  scale_fill_discrete(labels=c("No", "Yes")) +
  xlab("Death status")+
  ylab("Number of patients")
```

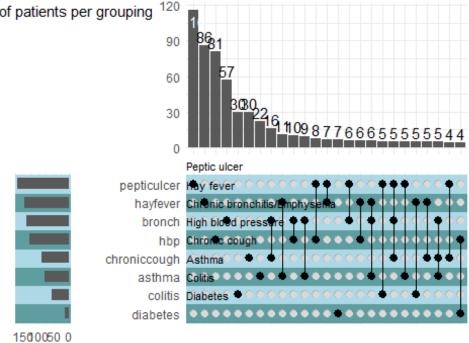


## Displaying the basis upset plot showing co-occurences of diseases in the dataset

```
############# Select codes on term frequency basis #########
# subdata[c(diseases, "smokeyrs", "age")] <-</pre>
lapply(subdata[c(diseases, "smokeyrs", "age")], as.numeric)
disease_table= as.data.frame(as.table(colSums(subdata[,diseases])))
disease_table=disease_table[order(disease_table$Freq,decreasing=TRUE),]
## Adeline: eventually extend num_code
## select the 8 most frequent diseases
frequent diseases=head(as.character(disease table$Var1),num code)
## convert diseases dummy variables to logical variables
subdata[diseases] <-subdata[diseases] == 1 ## converse dummy variables for</pre>
diseases to logical values
## Mapping diseases variables with real world names
mapping_diseases <- data.frame(var=c(</pre>
"hbp", "diabetes", "pepticulcer", "hayfever", "bronch", "chroniccough", "asthma", "c
olitis", "hf", "hepatitis"), real_names=c("High blood pressure", "Diabetes",
"Peptic ulcer", "Hay fever", "Chronic bronchitis/Emphysema", "Chronic cough",
"Asthma", "Colitis", "Heart failure", "Hepatitis"))
############ Essential upset plot
plottitle="Co-occurrence of diagnoses in patients"
plotsubtitle=paste("Diagnoses are selected by term frequency . Patient pool:
,nrow(subdata)," individuals.",sep="")
u=NULL
u=upset(
    data=subdata,
    intersect=frequent diseases,
    name="Co-occurence diagnoses",
   mode="exclusive intersection",
   min_size = min_size,
    n_intersections=num_comb,
    keep_empty_groups=FALSE,
    min degree=min degree,
    height_ratio=c(0.8,0.2),
    set size=FALSE,
```

```
matrix=(
        intersection_matrix(
            geom=geom_point(shape='circle',size=2),
            segment=geom segment(alpha=0.4)
         + annotate(
            geom='text',
            color="black",
label=mapping diseases[match(frequent diseases, mapping diseases[, "var"]), "rea
1_names"],
            x=-Inf,
            y=frequent_diseases,
            size=3,
            vjust=-1.5,
            hjust=0
        )+coord_cartesian(clip = "off")
+labs(title=plottitle,subtitle=plotsubtitle)+theme(plot.title=element_text(fa
ce="bold"))
d=length(unique(as.numeric(u$data$group)))
# get max y in the upset plot frequencies
temp=as.data.frame(table(u$patches[[1]][[1]]$data$exclusive_intersection))
my=max(temp[temp$Var1!="Outside of known sets",2])
#u
upset(
    data=subdata,
    intersect=frequent_diseases,
    name=paste("Diagnoses"," groupings by frequency",sep=""),
    mode="exclusive_intersection",
    min_size = min_size,
    keep empty groups=FALSE,
    min_degree=min_degree,
    height_ratio=c(1,0.1),
    width_ratio=c(0.1,0.5),
    n_intersections=num_comb,
    set_size=upset_set_size()+ ylab('n. of patients per code'),
    encode sets=FALSE,
    base annotations=list(
        'Intersection
```

```
size'=intersection_size(text_colors=c(on_background='black',on_bar='white'))
        + ylab('')
        + annotate("text", x=-Inf, y=my, label = "n. of patients per grouping",
hjust=1.2)
        + coord cartesian(clip = "off")
    stripes=upset_stripes(mapping = aes(),geom=geom_segment(size = 7),colors
= c("cadetblue", "lightblue"),data = NULL),
    matrix=(
        intersection_matrix(
            geom=geom_point(shape='circle', size=2),
            segment=geom_segment(alpha=0.4)
        )+ annotate(
            geom='text',
            color="black",
label=mapping diseases[match(frequent diseases, mapping diseases[, "var"]), "rea
1 names"],
            #label=frequent_diseases,
            x=-Inf,
            y=frequent_diseases,
            size=3,
            vjust=-1.5,
            hjust=0
        )+coord_cartesian(clip = "off")))
of patients per grouping
```



Diagnoses groupings by frequency

patients per code

### Comment

We can see from this plot that most of the patients are suffering only from one of these diseases in decreasing order: peptic ulcer, hay fever, High blood pressure, Chronic bronchitis/Emphysema. In general simultaneous occurrence of diseases seem to be: chronic bronchitis and chronic cough; Hay fever and asthma; chronic bronchitis and high blood pressure; chronic bronchitis and asthma. In our dataset, some patients are also suffering from three diseases simultaneously. These are: Hay fever, chronic bronchitis and asthma; Peptic ulcer, chronic bronchitis and chronic cough. Depending on the dataset content and structure, even more combinations of co-occurrence of diagnoses can be shown. This way a better quantified visualization of simultaneous co-occurrence of diagnoses his made possible and facilitates the general overview. Another interesting perspective would now to know the characteristics of those patients in terms for instance of gender, age, education, etc. This is done in the next section.

A pdf file of this plot has been made available in the folder under the name upset\_plot\_base for a better appreciation of the graph.

### Annotating the upset plot with other features to see patterns

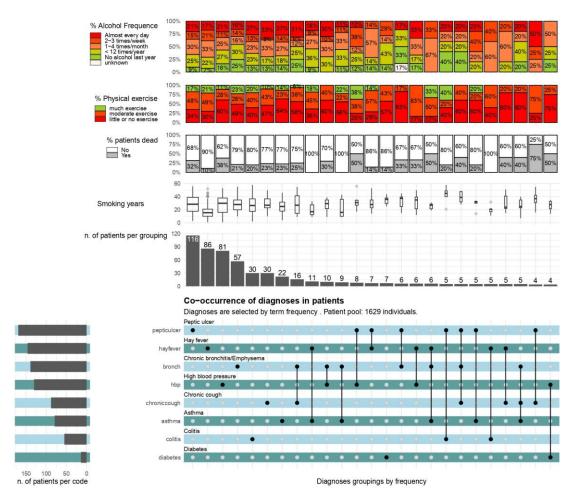
```
alcohol freq values=c(
# 0: Almost every day, 1: 2-3 times/week, 2: 1-4 times/month, 3: < 12
times/year, 4: No alcohol last year, 5: Unknown
"0" ="red",
        "orangered",
"1"=
"2"=
        "sienna1",
        "yellow3",
      "yellowgreen",
"4"=
"5"=
        "seashell"
exercise_freq_values=c(
# 0:much exercise,1:moderate exercise,2:little or no exercise
"0"=
        "yellowgreen",
        "orangered",
"1"=
"2"=
       "red"
)
# titles, captions
pdfname="co-occurrence of diagnoses in patients.pdf"
# Annotated upset plot
g=NULL
```

```
g=upset(
    data=subdata,
    intersect=frequent_diseases,
    name=paste("Diagnoses"," groupings by frequency", sep=""),
    mode="exclusive_intersection",
    min_size = min_size,
    keep_empty_groups=FALSE,
    min_degree=min_degree,
    height ratio=c(1,0.1),
    width_ratio=c(0.1,0.5),
    n intersections=num comb,
    set_size=upset_set_size()+ ylab('n. of patients per code'),
    encode_sets=FALSE,
    base_annotations=list(
        'Intersection
size'=intersection_size(text_colors=c(on_background='black',on_bar='white'))
        + ylab('')
        + annotate("text", x=-Inf, y=my, label = "n. of patients per grouping",
        + coord cartesian(clip = "off")
    stripes=upset_stripes(mapping = aes(),geom=geom_segment(size = 7),colors
= c("cadetblue", "lightblue"),data = NULL),
    matrix=(
        intersection_matrix(
            geom=geom_point(shape='circle', size=2),
            segment=geom_segment(alpha=0.4)
         + annotate(
            geom='text',
            color="black",
label=mapping_diseases[match(frequent_diseases, mapping_diseases[, "var"]), "rea
1_{\text{names}}
            x=-Inf,
            y=frequent diseases,
            size=3,
            vjust=-1.5,
            hjust=0
        )+coord_cartesian(clip = "off")
    ),
    annotations = list(
    'Alcohol Frequence'=ggplot(mapping=aes(fill=alcoholfreq))
        + geom_bar(stat='count', position='fill',color="black")
        + scale_y_continuous(labels=scales::percent_format())
        + scale_fill_manual(labels = c("Almost every day", "2-3 times/week",
"1-4 times/month", "< 12 times/year", "No alcohol last year", "unknown"),
```

```
values=alcohol_freq_values)
         + ylab('')
        + labs(fill='% Alcohol Frequence')
        + geom text(
            aes(label=!!aes percentage(relative to='intersection')),
            stat='count',
            size=3,
            position=position_fill(vjust = 0.5)
    #qeom text(aes(label=ifelse(percent >= 0.07, paste0(sprintf("%.0f",
percent*100), "%"), "")),
               # position=position_stack(vjust=0.5), colour="white")
        + theme(
            legend.position = c(-0.06, 1),
            legend.justification = c("right", "top"),
            legend.direction="vertical",
            legend.key.height=unit(0.1,"cm"),
            axis.title.y = element_text(angle = 0, vjust = 0.5)
             ),
'exercise'=ggplot(mapping=aes(fill=exercise))
        + geom_bar(stat='count', position='fill',color="black")
        + scale y continuous(labels=scales::percent format())
        + scale_fill_manual( labels = c("much exercise", "moderate
exercise","little or no exercise"), values=exercise freq values)
        + ylab('')
        + labs(fill="% Physical exercise")
        + geom text(
            aes(label=!!aes percentage(relative to='intersection')),
            stat='count',
            size=3,
                  position=position fill(vjust = 0.5)
        + theme(
            legend.position = c(-0.06, 0.8),
            legend.justification = c("right", "top"),
            legend.direction="vertical",
            legend.key.height=unit(0.1,"cm"),
            axis.title.y = element_text(angle = 0, vjust = 0.5)
             ),
    'patient dead'=ggplot(mapping=aes(fill=death))
        + geom_bar(stat='count', position='fill',color="black")
      # + scale fill discrete(name= '% patients dead', labels = c("No",
"Yes"))
        + scale y continuous(labels=scales::percent format())
        + scale_fill_manual(labels = c("No", "Yes"), values=c('white', 'grey'))
        + ylab('')
        + labs(fill='% patients dead')
        + geom text(
```

```
aes(label=!!aes_percentage(relative_to='intersection')),
            stat='count',
            size=3,
                 position=position fill(vjust = 0.5)
        + theme(
            legend.position = c(-0.06, 1),
            legend.justification = c("right", "top"),
            legend.direction="vertical",
            legend.key.height=unit(0.1, "cm"),
            axis.title.y = element_text(angle = 0, vjust = 0.5)
             ),
'Smoking years'=ggplot(mapping=aes(y=smokeyrs))
        + geom_boxplot(#stat='count', position='fill',color="black"
          varwidth = TRUE, alpha=0.2 )
         + ylab('Smoking years')
          + theme(
            legend.position = "none",
            #plot.title = element_text(hjust = -0.3),
           # plot.margin = rep(grid::unit(0.75, "in"),4)
           axis.title.y = element_text(angle = 0, vjust = 0.5)
           )
        )
+labs(title=plottitle,subtitle=plotsubtitle)+theme(plot.title=element_text(fa
ce="bold"))
# relative proportions of the various parts of the plot
h=1
k=0.5
alcohol_h=1.5
exercise h=1.1
dead_h=1.1
smoking h=1.1
his h=1.5
mat x=0.5
# upper part including histogram...
s_plots=(alcohol_h+exercise_h+dead_h+smoking_h+his_h)*h
# combinations panel...
s table=d*k
# ...altogether
plot_space=s_plots+s_table
# final plot
g=g+plot_layout(heights=c(h*alcohol_h,h*exercise_h,h*dead_h,h*smoking_h,h*his
h,k*d),ncol=2)
# Saving as a pdf file
# pdf("diagnoses_upset_plot.pdf", width=12, height=plot_space)
```

```
#g
dev.off()
## null device
## 1
knitr::include_graphics("./diagnoses_upset_plot.jpg")
```



### **Comment**

#### NB:

This last visualization is made available on pdf and jpg file formats in the folder under the name diagnoses\_upset\_plot for a better appreciation of the graph as a whole.

Just as said above, an interesting step when displaying co-occurrence of diagnoses in a patient dataset is to see the patterns of those patients. Some of those features might be age, gender, education, race, smoking status, death status, etc, . In the plot above, I choose to display: frequency of drinking alcohol, frequency of practicing a physical exercise, years of smoking and death status. So out of 1629 patients in our pool, 116 were suffering only from Peptic ulcer. And in this subset of patients suffering uniquely from peptic ulcer, they were smokers for more than 20 years, more than 80% were no, little or moderate active in terms

of physical exercise. Moreover 32% of them died. Similar comments can be drawn for patients diagnoses simultaneously of 2 or more diseases. For instance, of the 16 patients affected simultaneously of chronic cough and chronic bronchitis, 25% died, more than 90% are not really frequently practicing a physical exercise, over 30% drink alcohol almost every day.