

COVID-19 Situation in Spain

► Libraries

↳ 4 cells hidden

► Data loading and general overview

First, we will load the csv file from the official Spanish government site which as we said is updated every 24 hours. The data can be automatically downloaded from the Spanish government website (updated every 24 hours). The data can be automatically downloaded from the Spanish government website (updated every 24 hours) and saved in our dataframe called data.

↳ 16 cells hidden

▼ Plotting

```
aux = total_s.melt(id_vars="Date", value_vars=("Cases", "Infected", "TestAc+", "Deaths", "ICU", "Hosp"))
aux.head()
```



	Date	Description	Count
0	2020-02-20	Cases	0.0
1	2020-02-21	Cases	0.0
2	2020-02-22	Cases	0.0
3	2020-02-23	Cases	0.0
4	2020-02-24	Cases	0.0

```
data_infected = data[data.Date>"20-02-2020"]
```

▼ Infections over time

```
fig = px.bar(data_infected, x="CCAA", y="Infected", color="CCAA",
              animation_frame="NEW_DATE", animation_group="CCAA", range_y=[0, data.Infected.max()])
fig.show()
```

```
daily.insert(0,0)
```

```
exec('df_{}'.format(a))
```

```
df_daily_infected = pd.DataFrame({"Date":data.Date.unique(),
                                   "Madrid":df_Madrid["Daily_infected"].values,
                                   "Cataluña":df_Cataluña["Daily_infected"].values,
                                   "Andalucía":df_Andalucía["Daily_infected"].values,
                                   "Castilla La Mancha":df_Castilla_La_Mancha["Daily_infected"].values,
                                   "Castilla y León":df_Castilla_y_León["Daily_infected"].values,
                                   "País Vasco":df_País_Vasco["Daily_infected"].values})
```

```
aux_i = df_daily_infected.melt(id_vars="Date", value_vars=("Madrid", "Cataluña", "Andalucía", "Castilla La Mancha", "Castilla y León", "País Vasco"))
```

▼ Daily infections in Top 6 countries

```
aux_1=aux_i[aux_i.Date>"20-02-2020"]
```

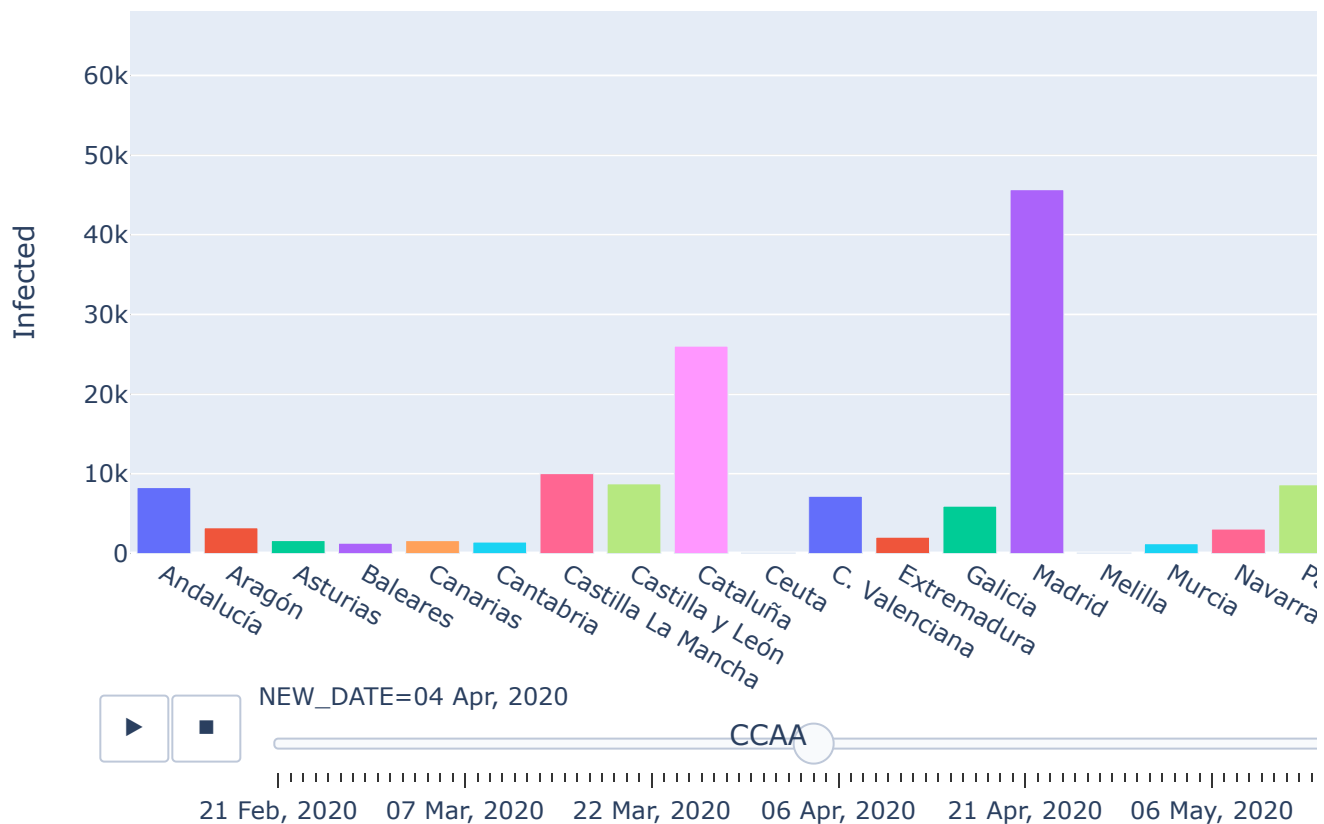
```
aux_2=aux_i[aux_i.Date>"18-04-2020"]
```

```
fig = px.bar (aux_2, x= "Date", y = "Count", color="CCAA", title= "Daily infections in Spain")
fig.show()
```





Infections by regions over time



▼ Analysis of Madrid

Madrid was the region with the highest number of infections and deaths. In the following graphics we will analyze the variables. The cases related to infections were collected in the variable PCR+ (Infected), we can just analyze the data for Madrid.

```
total_madrid = data[data.CCAA=="Madrid"].groupby("Date")["Date","Infected","Deaths","Hospitalized"]
```

```
aux_m = total_madrid.melt(id_vars="Date", value_vars=("Infected","Deaths","ICU","Hospitalized"))
```

```
fig = px.bar(aux_m, x= "Date", y = "Count", color="Status", title= "Actual situation in Madrid")
fig.show()
```



```

    'GA': 'Galicia',
    'MD': 'Madrid',
    'ML': 'Melilla',
    'MC': 'Murcia',
    'NC': 'Navarra',
    'PV': 'País Vasco',
    'RI': 'La Rioja'
}

```

```

d_ccaa = {
    'Andalucía': 'Andalucía',
    'Aragón': 'Aragón',
    'Asturias': 'Principado de Asturias',
    'Balears': 'Islas Baleares',
    'Canarias': 'Islas Canarias',
    'Cantabria': 'Cantabria',
    'Castilla La Mancha': 'Castilla-La Mancha',
    'Castilla y León': 'Castilla y León',
    'Cataluña': 'Cataluña',
    'Ceuta': 'Ceuta y Melilla',
    'C. Valenciana': 'Comunidad Valenciana',
    'Extremadura': 'Extremadura',
    'Galicia': 'Galicia',
    'Madrid': 'Comunidad de Madrid',
    'Melilla': 'Ceuta y Melilla',
    'Murcia': 'Región de Murcia',
    'Navarra': 'Comunidad Foral de Navarra',
    'País Vasco': 'País Vasco',
    'La Rioja': 'La Rioja'
}

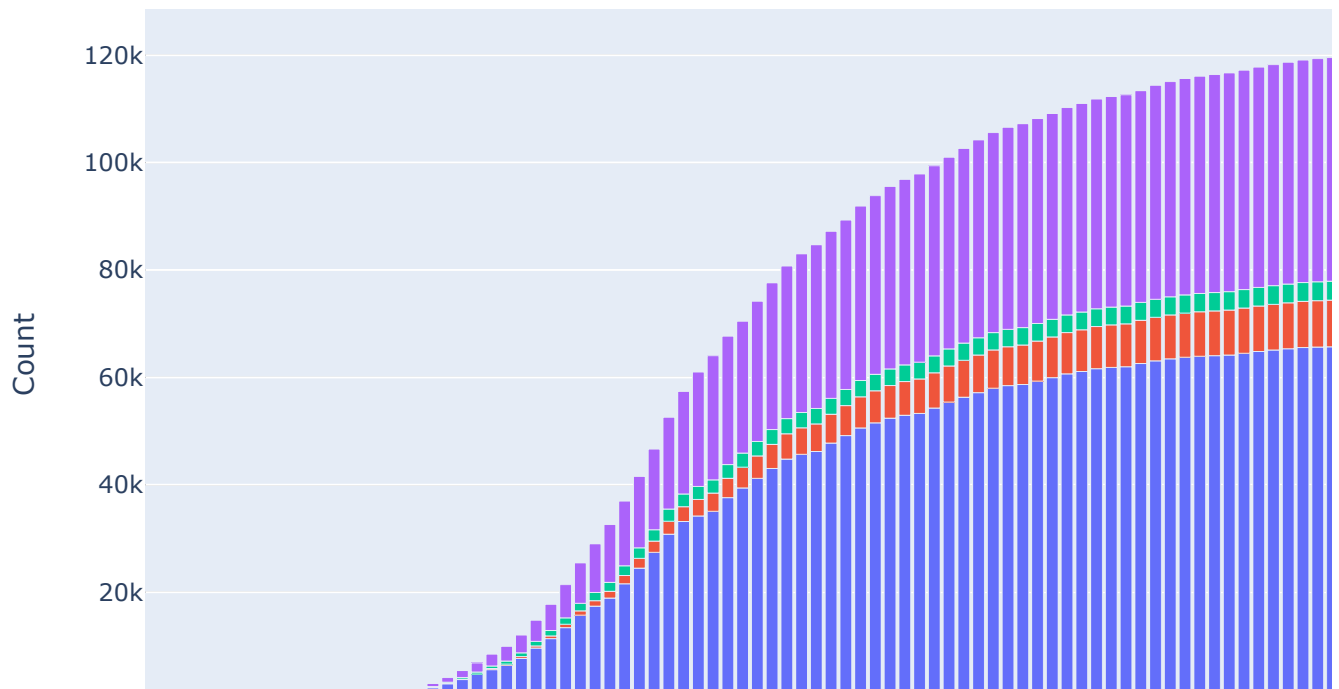
```

```

d_ccaa_id = {
    'Andalucía': "1",
    'Aragón' : "2",
    'Principado de Asturias': "3",
    'Islas Baleares': "4",
    'Islas Canarias': "5",
    'Cantabria': "6",
    'Castilla-La Mancha': "7",
    'Castilla y León': "8",
    'Cataluña': "9",
    'Ceuta y Melilla': "10",
    'Comunidad Valenciana': "11",
    'Extremadura': "12",
    'Galicia': "13",
    'Comunidad de Madrid' : "14",
    'Ceuta y Melilla': "15",
    'Región de Murcia': "16",
    'Comunidad Foral de Navarra': "17",
    'País Vasco': "18",
}

```

Actual situation in Madrid



```
for i in data.CCAA.unique():

    a = i.replace(".", "")
    a = a.replace(" ", "_")

    exec('df_{}=data[data.CCAA == i]'.format(a))

    exec('aux_a = df_{}.Infected.to_list()'.format(a))

    daily=[]
    for i in range(len(aux_a)-1):
        b = aux_a[i+1] - aux_a[i]
        daily.append(b)

    daily.insert(0,0)

    exec('df_{}["Daily_infected"] = daily'.format(a))

    exec('aux_d = df_{}.Deaths.to_list()'.format(a))

    daily=[]
    for i in range(len(aux_d)-1):
        b = aux_d[i+1] - aux_d[i]
        daily.append(b)
```

```
'La Rioja': "19"
}
```

```
d_ccaa_population = {
'Andalucía': 8414240,
'Aragón' : 1319291,
'Principado de Asturias': 1022800,
'Islas Baleares': 1149460,
'Islas Canarias': 2153389,
'Cantabria': 581078,
'Castilla-La Mancha': 2032863,
'Castilla y León': 2399548,
'Cataluña': 7675217,
'Ceuta y Melilla': 171264,
'Comunidad Valenciana': 5003769,
'Extremadura': 1067710,
'Galicia': 2699499,
'Comunidad de Madrid' : 6663394,
'Ceuta y Melilla': 171264,
'Región de Murcia': 1493898,
'Comunidad Foral de Navarra': 654214,
'País Vasco': 2207776,
'La Rioja': 316798
}
```

```
def get_hex_colors(df, data_to_color, cmap = matplotlib.cm.Reds, log = False):
```

```
...
```

This function takes the following arguments

1. df:pandas DataFrame with the data.
2. data_to_color: the column name with data based on which we want to create the color
3. cmap: colors you want to plot. You can use this to communicate different messages.
default is matplotlib.cm.Reds
more about colormaps: [https://matplotlib.org/3.1.1/gallery/color/colormap_ref](https://matplotlib.org/3.1.1/gallery/color/colormap_reference.html)
3. log: if data has huge outliers, we can create the color map with a logarithmic norm
default is False.

```
...
```

```
cmap = cmap # define the color pallete you want. You can use Reds, Blues, Greens etc
my_values = df[data_to_color] # get the value you wan to convert to colors
```

```
mini = min(my_values) # get the min to normalize
maxi= max(my_values) # get the max to normalize
```

```
LOGMIN = 0.01 # arbitrary lower bound for log scale
```

```
if log:
```

```
    norm = matplotlib.colors.LogNorm(vmin=max(mini,LOGMIN), vmax=maxi) # normalize log da
```

```

else:
    norm = matplotlib.colors.Normalize(vmin=mini, vmax=maxi) # create a color range

    colors = {value:matplotlib.colors.rgb2hex(cmap(norm(value))[:3]) for value in sorted(list

return colors

def get_hex_colors_2(value, cats):
    '''
    Color paletter used from this website:

    https://colorbrewer2.org/#type=sequential&scheme=Reds&n=9

    The color selection will be based on the percentile each value is in.
    '''
    if value == 0:
        return "#FFFFFF"
    elif value in cats[0]:
        return "#fff5f0"
    elif value in cats[1]:
        return "#fee0d2"
    elif value in cats[2]:
        return "#fcbba1"
    elif value in cats[3]:
        return "#fc9272"
    elif value in cats[4]:
        return "#fb6a4a"
    elif value in cats[5]:
        return "#ef3b2c"
    elif value in cats[6]:
        return "#cb181d"
    elif value in cats[7]:
        return "#a50f15"
    elif value in cats[8]:
        return "#67000d"
    else:
        return "#000000"

df = pd.read_csv("https://covid19.isciii.es/resources/serie\_historica\_acumulados.csv",delimit

df.rename(columns = {"FECHA":"DATE",
                    "CASOS":"CASES",
                    "PCR+": "TOTAL_INFECTED",
                    "Hospitalizados":"REQUIERED_HOSPITALIZATION",
                    "UCI":"REQUIERED_ADVANCED_CARE",
                    "Fallecidos":"TOTAL_DEATHS"}, inplace = True)

df.fillna(0, inplace = True)
df["CCAA"] = df["CCAA"].map(d name)

```

Daily infections in Spain (Top 6)

▼ Daily deaths in Top 6 countries

```

df_daily_fatalities = pd.DataFrame({"Date":data.Date.unique(),
                                    "Madrid":df_Madrid["Daily_deaths"].values,
                                    "Cataluña":df_Cataluña["Daily_deaths"].values,
                                    "Valencia":df_C_Valenciana["Daily_deaths"].values,
                                    "Castilla La Mancha":df_Castilla_La_Mancha["Daily_deaths"].values,
                                    "Castilla y Leon":df_Castilla_y_León["Daily_deaths"].values,
                                    "País Vasco":df_País_Vasco["Daily_deaths"].values})

aux_f = df_daily_fatalities.melt(id_vars="Date", value_vars=("Madrid","Cataluña","Valencia",'
fig = px.line (aux_f, x= "Date", y = "Count", color="CCAA", title= "Daily Death in Spain (Top
fig.show()
```




```

data_to_color = "TOTAL_DEATHS"
cats, bins = pd.qcut(df[data_to_color].unique()[np.argsort(df[data_to_color].unique())], q =
cats = cats.unique()

#-----

# value we will iterate in order to create the styledict
ccaas = list(df["id"].unique())
dates = list(df["DATE_for_Folium"].unique())

# create the color dict and color column
df["COLORS"] = df[data_to_color].apply(get_hex_colors_2, args = [cats]) # we create a column i

# creates the styledict for the map
styledict = {}

# iterate the populate the styledict
for ccaa in ccaas:
    styledict[str(ccaa)] = {date: {'color': df[(df["id"] == ccaa) & (df["DATE_for_Folium"] ==
                                'opacity': 0.6} for date in dates)}

# creates and renders the Folium map
m = folium.Map(location=(40, 0), tiles='OpenStreetMap', zoom_start=6)

g = TimeSliderChoropleth(
    gdf.set_index("id").to_json(), # get's the coordinates for each id
    styledict = styledict # styledict contains for each id the timestamp and the color to plac
)

m.add_child(g)

#-----
# Let's create a legend for folium
# https://nbviewer.jupyter.org/gist/talbertc-usgs/18f8901fc98f109f2b71156cf3ac81cd

from branca.element import Template, MacroElement

template = """
{% macro html(this, kwargs) %}

<!doctype html>
<html lang="en">
<head>
  <meta charset="utf-8">
  <meta name="viewport" content="width=device-width, initial-scale=1">
  <title>jQuery UI Draggable - Default functionality</title>
  <link rel="stylesheet" href="//code.jquery.com/ui/1.12.1/themes/base/jquery-ui.css">

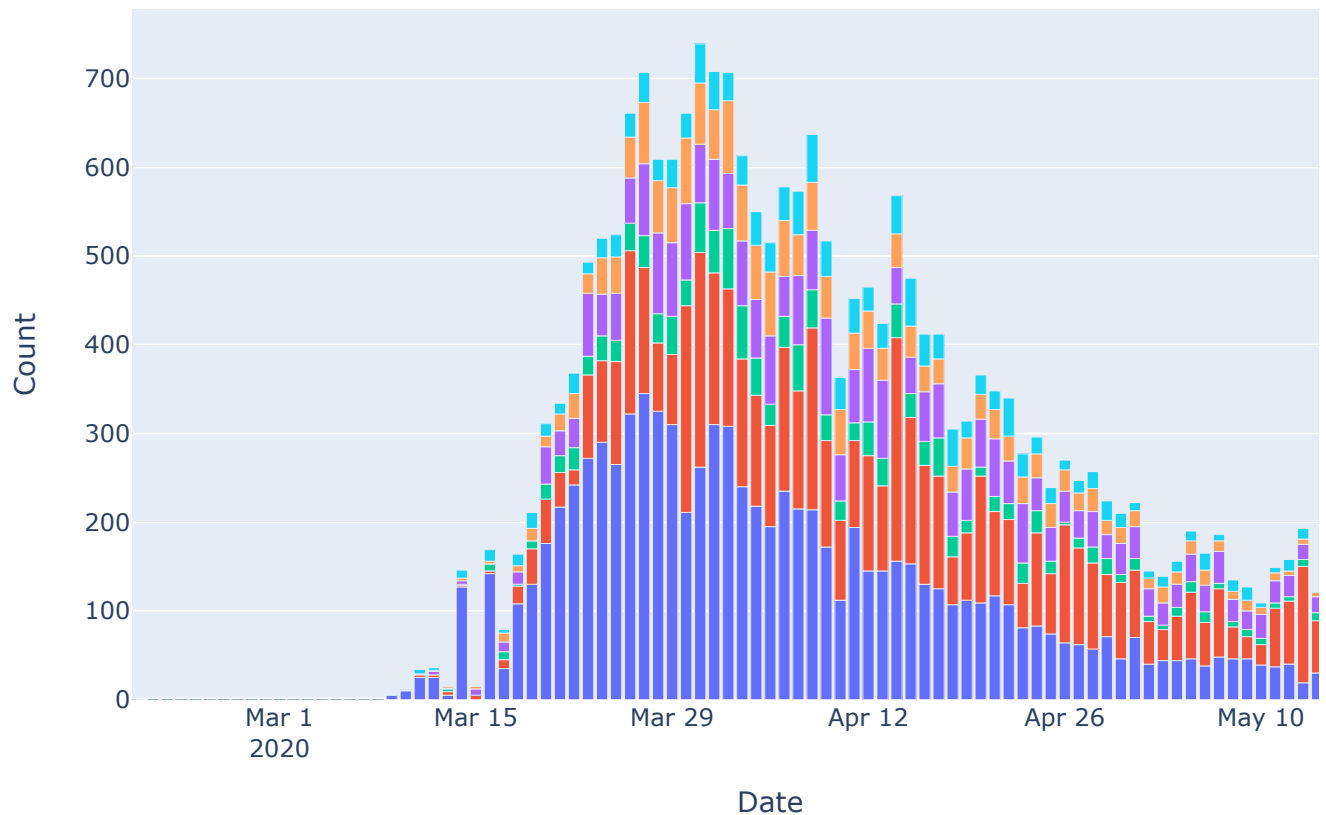
  <script src="https://code.jquery.com/jquery-1.12.4.js"></script>
  <script src="https://code.jquery.com/ui/1.12.1/jquery-ui.js"></script>

```

```
fig = px.bar (aux_f, x= "Date", y = "Count", color="CCAA", title= "Daily Deaths in Spain (Top 6)"
fig.show()
```



Daily Deaths in Spain (Top 6)



▼ Mapping with choropleth

```
d_name = {
'AN': 'Andalucía',
'AR': 'Aragón',
'AS': 'Asturias',
'IB': 'Balears',
'CN': 'Canarias',
'CB': 'Cantabria',
'CM': 'Castilla La Mancha',
'CL': 'Castilla y León',
'CT': 'Cataluña',
'CE': 'Ceuta',
'VC': 'C. Valenciana',
'EX': 'Extremadura',
'GA': 'Galicia'
```

```

df["CCAA_for_Folium"] = df["CCAA"].map(d_ccaa)
df["id"] = df["CCAA_for_Folium"].map(d_ccaa_id)

df["Population"] = df["CCAA_for_Folium"].map(d_ccaa_population)

df["CCAA"].isnull().sum()

```

 0

```

def correct_date(date_str):
    list_dates = date_str.split("/")
    day = list_dates[0]
    month = list_dates[1]
    year = list_dates[2]

    if len(day) == 1:
        day = "0" + day
    if len(month) == 1:
        month = "0" + month

    return "/".join([day, month, year])

df["NEW_DATE"] = df["DATE"].apply(correct_date)

df["DATE"] = pd.to_datetime(df["NEW_DATE"], format='%d/%m/%Y')

df["DATE_for_Folium"] = (df["DATE"].astype(int)// 10**9).astype('U10')

df = df[["id", "CCAA", "CCAA_for_Folium", "DATE", "DATE_for_Folium", "TOTAL_INFECTED", "REQUI
df["id"].astype(np.int16)
df.head()

PATH_GEO_JSON = 'shapefiles_ccaa_espana.geojson'
gdf = gpd.read_file(PATH_GEO_JSON)
gdf["id"] = gdf["name_1"].map(d_ccaa_id) # create a numerical id for each ccaa
gdf = gdf[["id", "shape_leng", "shape_area", "geometry"]] # extract the id and the geometry (cc
gdf["geometry"] = gdf["geometry"].simplify(0.1, preserve_topology = False)
gdf["id"].astype(int)
gdf.head()

```

▼ Infections

```
m = folium.Map(location = (40, 0), zoom_start = 5.5)
```

```

<script>
$( function() {
    $( "#maplegend" ).draggable({
        start: function (event, ui) {
            $(this).css({
                right: "auto",
                top: "auto",
                bottom: "auto"
            });
        }
    });
});

</script>
</head>
<body>

<div id='maplegend' class='maplegend'
    style='position: absolute; z-index:9999; border:2px solid grey; background-color:rgba(255,255,255,0.6); border-radius:6px; padding: 10px; font-size:14px; right: 20px; bottom: 20px;'>

<div class='legend-title'>Legend</div>
<div class='legend-scale'>
    <ul class='legend-labels'>
        <li><span style='background:#FFFFFF;opacity:0.6;'></span>No cases</li>
        <li><span style='background:#fff5f0;opacity:0.6;'></span>1 Quantile</li>
        <li><span style='background:#fee0d2;opacity:0.6;'></span>2 Quantile</li>
        <li><span style='background:#fcbba1;opacity:0.6;'></span>3 Quantile</li>
        <li><span style='background:#fc9272;opacity:0.6;'></span>4 Quantile</li>
        <li><span style='background:#fb6a4a;opacity:0.6;'></span>5 Quantile</li>
        <li><span style='background:#ef3b2c;opacity:0.6;'></span>6 Quantile</li>
        <li><span style='background:#cb181d;opacity:0.6;'></span>7 Quantile</li>
        <li><span style='background:#a50f15;opacity:0.6;'></span>8 Quantile</li>
        <li><span style='background:#67000d;opacity:0.6;'></span>9 Quantile</li>
        <li><span style='background:#000000;opacity:0.6;'></span>Other</li>
    </ul>
</div>
</div>

</body>
</html>

<style type='text/css'>
.maplegend .legend-title {
    text-align: left;
    margin-bottom: 5px;
    font-weight: bold;
    font-size: 90%;
}
.maplegend .legend-scale ul {

```

```

margin: 0;
margin-bottom: 5px;
padding: 0;
float: left;
list-style: none;
}
.maplegend .legend-scale ul li {
font-size: 80%;
list-style: none;
margin-left: 0;
line-height: 18px;
margin-bottom: 2px;
}
.maplegend ul.legend-labels li span {
display: block;
float: left;
height: 16px;
width: 30px;
margin-right: 5px;
margin-left: 0;
border: 1px solid #999;
}
.maplegend .legend-source {
font-size: 80%;
color: #777;
clear: both;
}
.maplegend a {
color: #777;
}
</style>
{% endmacro %}"""

```

```

macro = MacroElement()
macro._template = Template(template)

```

```
m.get_root().add_child(macro)
```

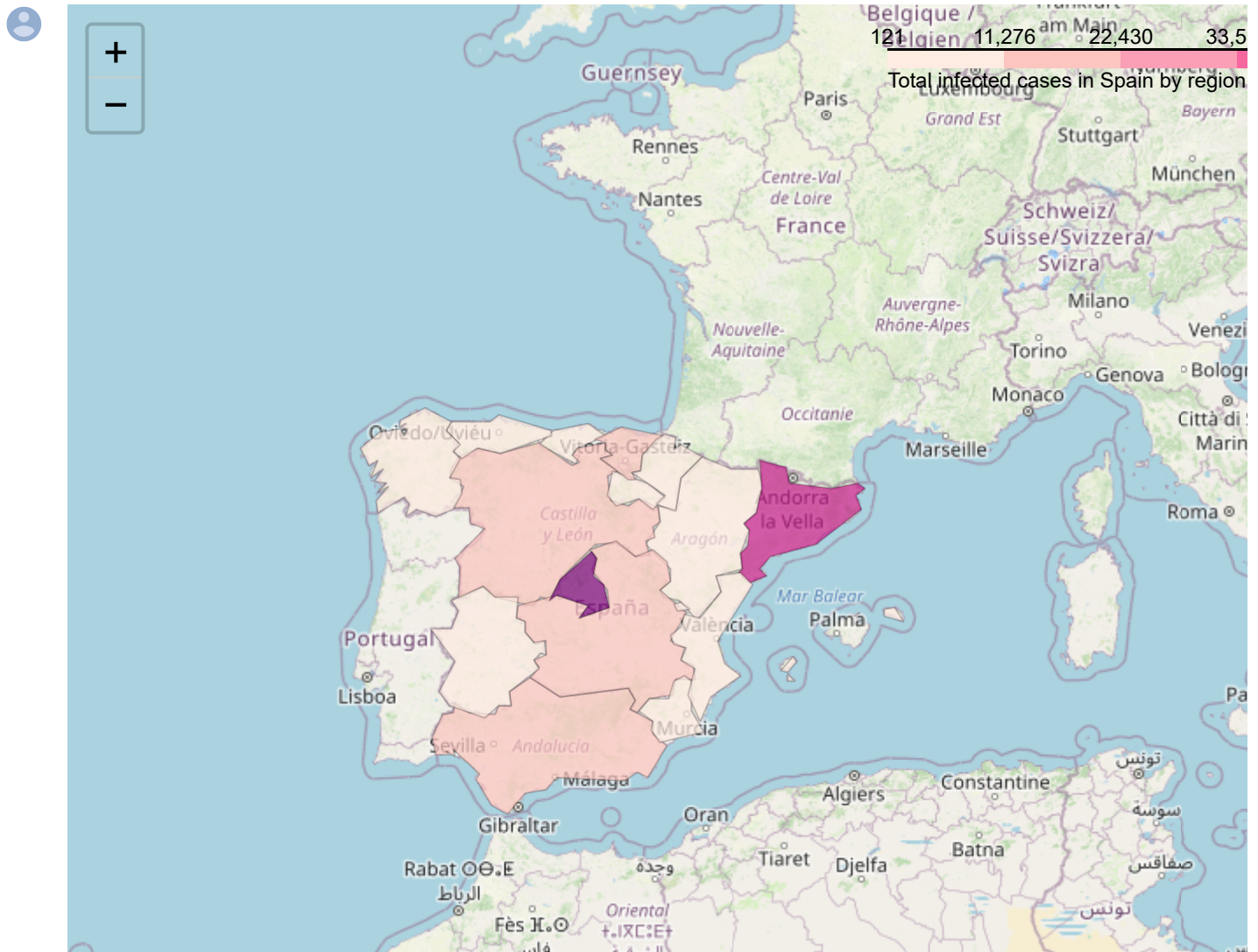


```

folium.Choropleth(
    geo_data = gdf,
    name = 'choropleth',
    data = df[df["DATE"] == max(df["DATE"])],
    columns = ['id', 'TOTAL_INFECTED'],
    key_on='feature.properties.id',
    fill_color='RdPu',
    fill_opacity=0.7,
    line_opacity=0.2,
    legend_name = 'Total infected cases in Spain by region'
).add_to(m)

```

m



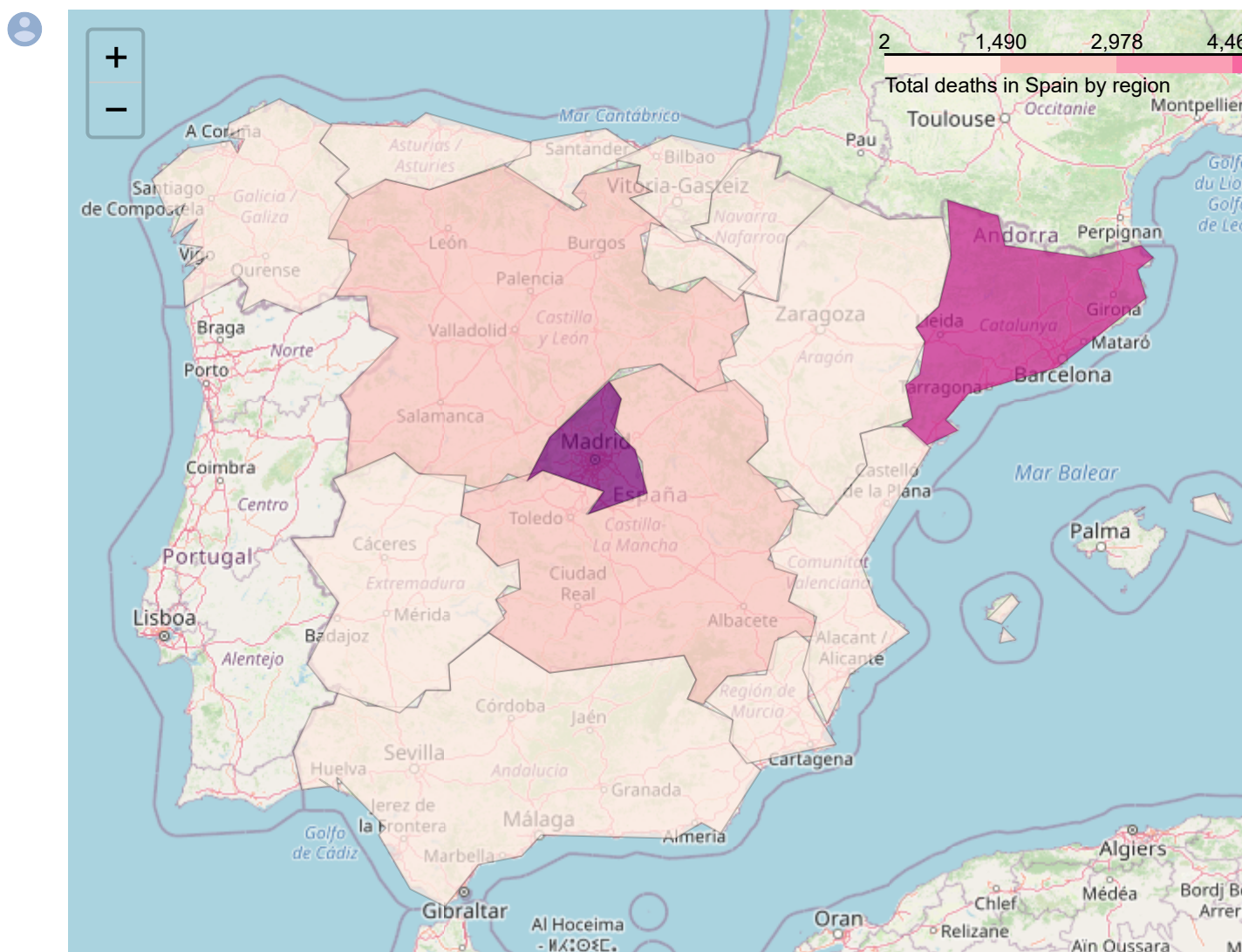
▼ Deaths

```
m = folium.Map(location = (40, -2), zoom_start = 5.5)
```

```
m = folium.Map(location = (40, 0), zoom_start = 5.5)
```

```
folium.Choropleth(
    geo_data = gdf,
    name = 'choropleth',
    data = df,
    columns = ['id', 'TOTAL_DEATHS'],
    key_on='feature.properties.id',
    fill_color='RdPu',
    fill_opacity=0.7,
    line_opacity=0.2,
    legend_name = 'Total deaths in Spain by region'
).add_to(m)
```

```
m
```



▼ Deaths time slider



▼ Mortality rate



```
df["Infected_1000h"] = df["TOTAL_INFECTED"]/(df["Population"]/1000)
df["Mortality_rate"] = df["TOTAL_DEATHS"] / df["TOTAL_INFECTED"]
df.fillna(0, inplace = True)
df.tail()
```



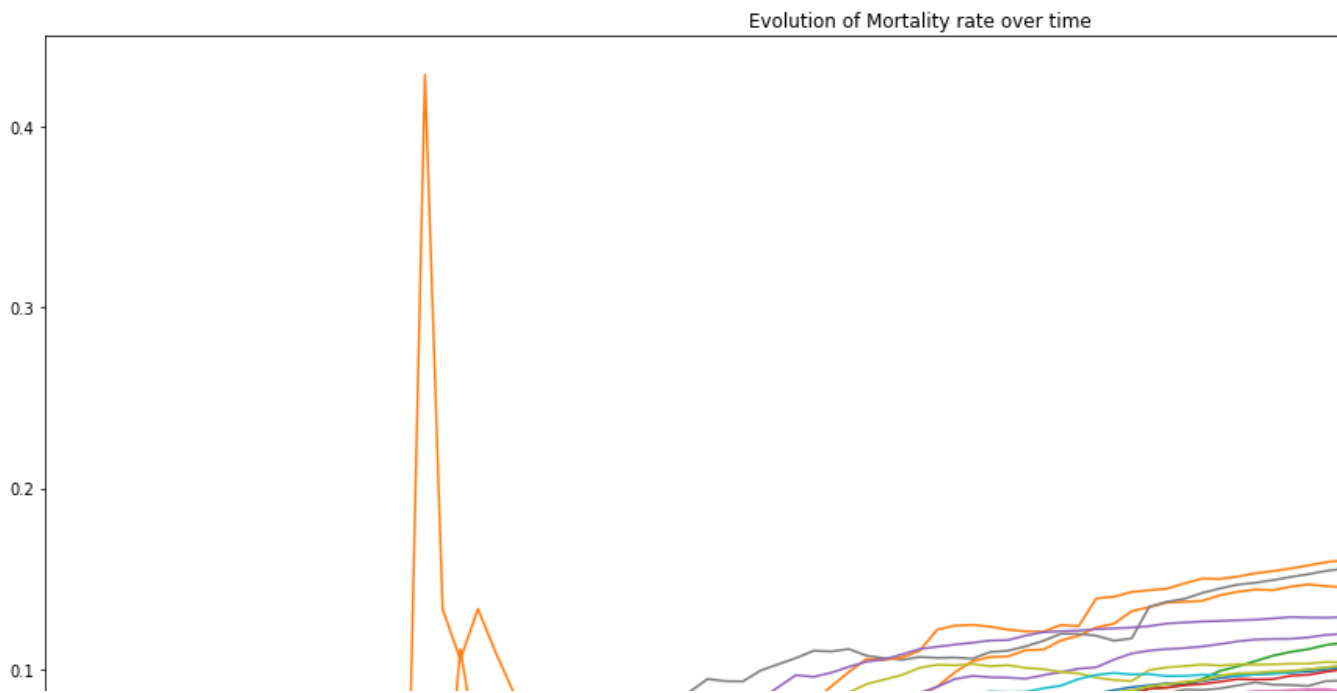
```
plt.figure(figsize = (20, 10))
```

```
for ccaa in sorted(list(df["CCAA"].unique())):
```

```
    x = df["DATE"].unique()
    y = df[df["CCAA"] == ccaa]["Mortality_rate"]
```

```
    plt.plot(x, y, label = ccaa)
    plt.title("Evolution of Mortality rate over time")
    plt.legend()
    plt.xticks(rotation=90)
```





```
x = [day for day in range(len(df["DATE"].unique()))]
```

```
fig, axes = plt.subplots(nrows=4, ncols=5, figsize=(30,20))
```

```
# plt.setp(axes, ylim=(0 ,max(df["Mortality rate"])))
```

```
ccaas = list(df["CCAA"].unique())
```

```
i = 0
```

```
for col_axes in axes:
```

```
    for ax in col_axes:
```

```
        if i < len(ccaas):
```

```
            ccaa = ccaas[i]
```

```
            y = df[df["CCAA"] == ccaa]["Mortality_rate"].values
```

```
            ipeaks, _ = find_peaks(y)
```

```
            ax.plot(x, y, color = "k", alpha = 0.7)
```

```
            ax.scatter(ipeaks, np.array(y)[ipeaks], color = "red", label = "Local peaks of mc
```

```
            ax.scatter(x[list(y).index(np.max(y))], np.max(y), color = "k", marker = "o", alp
```

```
            ax.set_title("Mortality rate {}".format(ccaa))
```

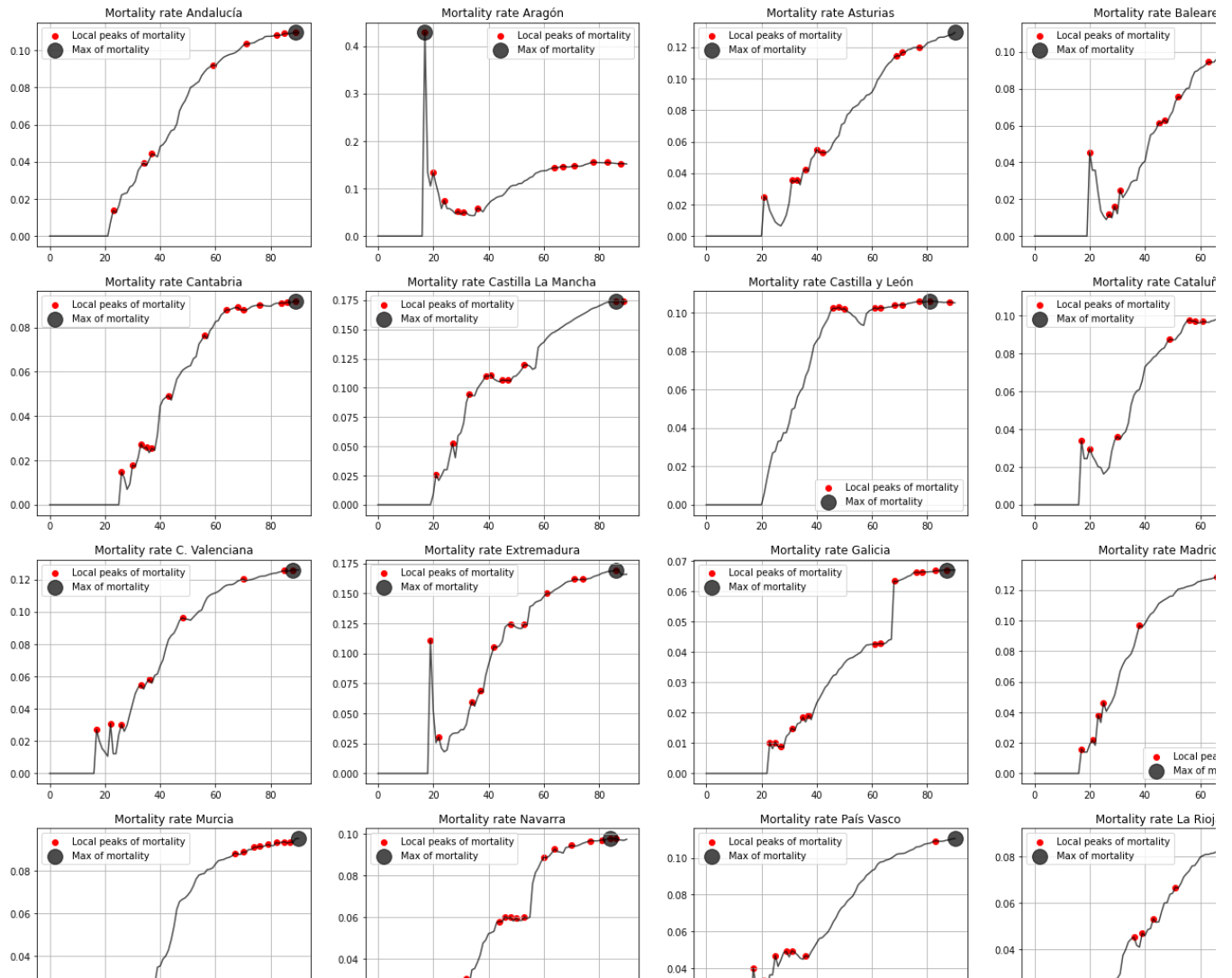
```
            ax.legend()
```

```
            ax.grid()
```

```
            i += 1
```

```
fig.delaxes(axes[3, 4])
```





create the pivot table of total cases

```
total_df = df.set_index("DATE").resample("D")[["TOTAL_INFECTED", "REQUIERED_HOSPITALIZATION",
total_df = total_df[total_df["Population"] > 0]
total_df["TOTAL_INFECTED_1000H"] = total_df["TOTAL_INFECTED"]/(total_df["Population"]/1000)
total_df["TOTAL_DEATHS_1000H"] = total_df["TOTAL_DEATHS"]/(total_df["Population"]/1000)
```

```
# get the data
x = list(total_df.index)
y_1 = list(total_df["TOTAL_INFECTED_1000H"]) # 1 axis
y_2 = list(total_df["TOTAL_DEATHS_1000H"]) # 2 axis
```

```
# create the figures
fig, ax = plt.subplots(figsize = (15, 7))
plot1 = ax.plot(x, y_1, color = "r", label = "Total infected per 1000 habitants") # plot the
plt.xticks(rotation=90) # rotate the date

ax2 = ax.twinx() # create a secondary axis
plot2 = ax2.plot(x, y_2, color = "k", label = "Total deaths per 1000 habitants") # plot the s
fig.tight_layout()
plt.title("Evolution of total infected cases and total deaths per 1000 habitants")
```

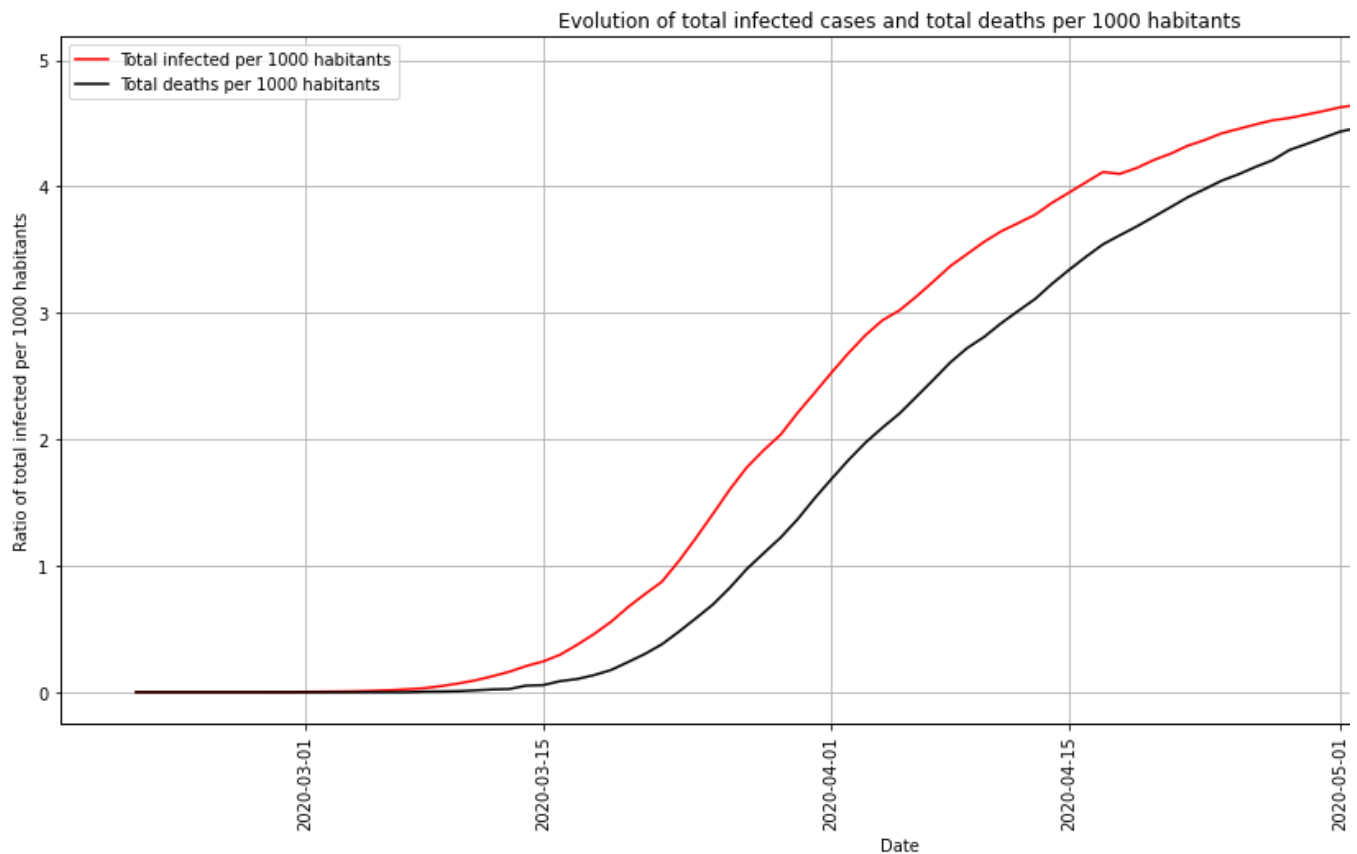
```
plt.plot(plt.gca().get_xdata(), total_infected_per_1000_habitants, color='red',
```

```
# create a common legend
lns = plot1 + plot2
labs = [l.get_label() for l in lns]
ax.legend(lns, labs, loc=0)

# prettify
ax.grid()
ax.set_xlabel("Date")
ax.set_ylabel("Ratio of total infected per 1000 habitants")
ax2.set_ylabel("Ratio of total deaths per 1000 habitants")
```



Text(998.875, 0.5, 'Ratio of total deaths per 1000 habitants')



▼ 8 of March manifestation

On 8 of March, the government allowed a massive manifestation all over the country for the international day of women. By that time, there were already some informed cases, but, we only know what we know. Now, since it takes 5 to 6 days to incubate in your body, this means that a person who was on the manifestation and began to have symptoms is likely that they were already infected or got infected. Let's make this shift in infected cases and see the results.

As you can see, when the manifestation was held, it is likely that there were around between 10k and 15k cases, which is more than the official 1006 cases.

```
total_df["SHIFT_7_DAYS"] = total_df["TOTAL_INFECTED"].shift(-7)
total_df["SHIFT_14_DAYS"] = total_df["TOTAL_INFECTED"].shift(-14)
```

▼ Known infected cases vs 'Real Cases' with a 1 and 2 week shift

```
x = np.array([x for x in range(len(total_df.index))])
y_informed = total_df["TOTAL_INFECTED"]
y_real_7_days = total_df["SHIFT_7_DAYS"]
y_real_14_days = total_df["SHIFT_14_DAYS"]
width = np.min(np.diff(x))/3

fig = plt.figure(figsize = (20, 10))

ax = fig.add_subplot(111)
ax.bar(x - width, y_informed, width, color = 'b', label = 'Known cases', alpha = 0.5)
ax.bar(x, y_real_7_days, width, color = 'r', label = '"Real Cases" shift 1 week', alpha = 0.4)
ax.bar(x + width, y_real_14_days, width, color='k', label = '"Real Cases" shift 2 week', alpha = 0.4)
ax.set_xlabel('Days since first infected case.')

plt.title("Known infected cases vs 'Real Cases' with a 1 and 2 week shift")
plt.axvline(x=17, lw = 1, alpha = 0.3, ymax = 0.4, color = "purple")
plt.annotate("8 March manifestation held", xy= (15, 80000), color = "purple")

textstr = '\n'.join((
    r'Known cases vs "Real Cases" 1 week shift: {:.0f}'.format(total_df.iloc[17]["SHIFT_7_DAYS"]),
    r'Known cases vs "Real Cases" 2 week shift: {:.0f}'.format(total_df.iloc[17]["SHIFT_14_DAYS"])
))

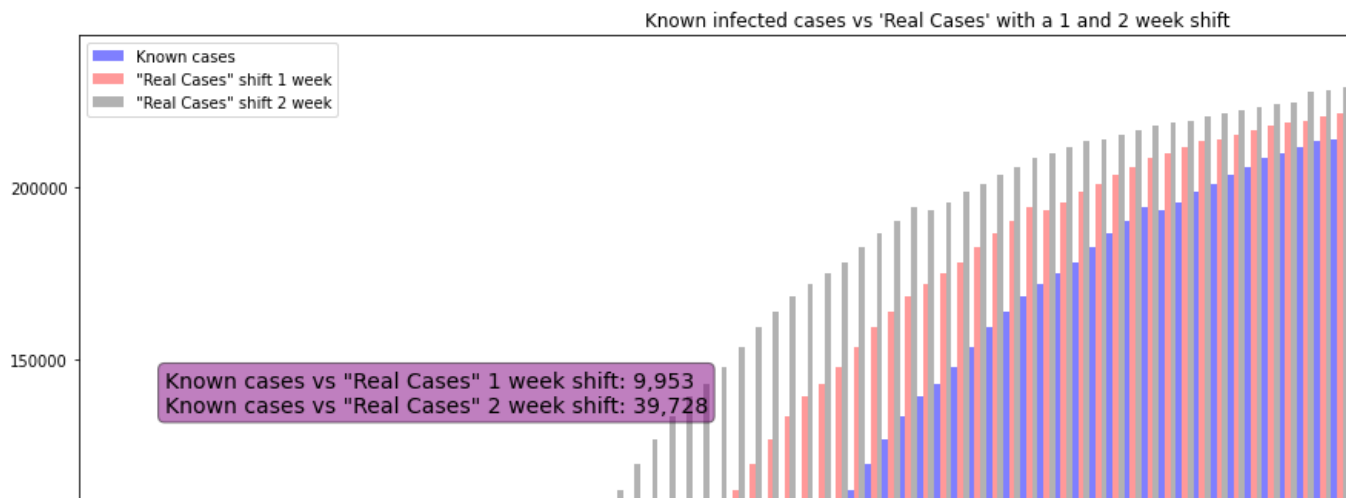
props = dict(boxstyle='round', facecolor='purple', alpha=0.5)

# place a text box in upper left in axes coords
ax.text(0.05, 0.6, textstr, transform=ax.transAxes, fontsize=14,
       verticalalignment='top', bbox=props)

plt.legend()
```



<matplotlib.legend.Legend at 0x7fa53a6eae8>



▼ Mortality rate by region

```
short_df = df[df["DATE"] == max(df["DATE"])]["CCAA", "Mortality_rate"].sort_values("Mortality_rate")
x = short_df["CCAA"]
y = short_df["Mortality_rate"]
```

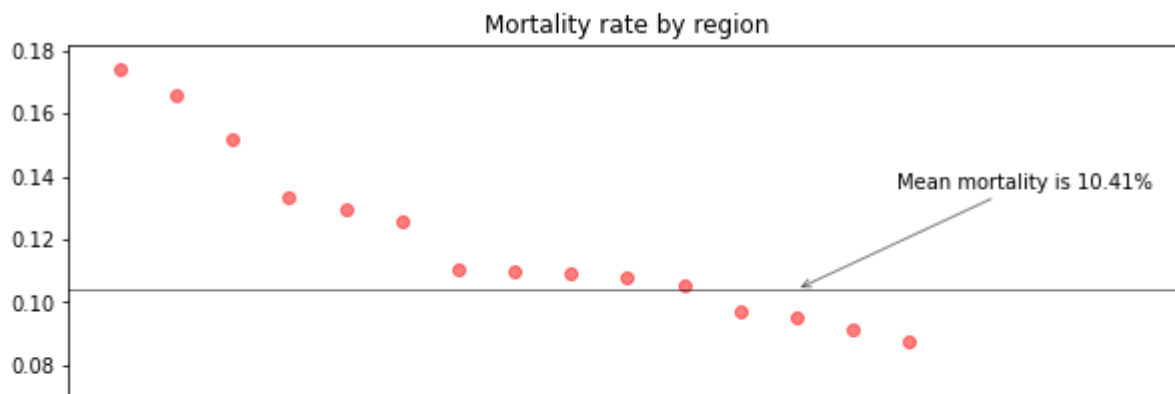
```
mean_y = np.mean(y)
mean_y
```

```
plt.figure(figsize = (10, 5))
plt.scatter(x, y, c = "red", alpha = 0.5)
plt.title("Mortality rate by region")
```

```
plt.xticks(rotation=90)
plt.axhline(mean_y, c = "k", alpha = 0.5, lw = 1)
plt.annotate('Mean mortality is {}%'.format(round(mean_y * 100, 2)),
            xy=(12, mean_y),
            xycoords='data',
            xytext=(50, 50),
            textcoords='offset points',
            arrowprops=dict(arrowstyle="->", color = "k", alpha = 0.5),
            color = "k")
```



```
Text(50, 50, 'Mean mortality is 10.41%')
```



▼ General Analysis over the time

▼ Data Load

```
infected = pd.read_csv('ccaa_covid19_casos_long.csv')
uci_beds = pd.read_csv('ccaa_camas_uci_2017.csv')
recovered = pd.read_csv('ccaa_covid19_altas_long.csv')
death = pd.read_csv('ccaa_covid19_fallecidos_long.csv')
hospitalized = pd.read_csv('ccaa_covid19_hospitalizados_long.csv')
masks = pd.read_csv('ccaa_covid19_mascarillas.csv')
uci = pd.read_csv('ccaa_covid19_uci_long.csv')
national = pd.read_csv('nacional_covid19.csv')
age_range = pd.read_csv('nacional_covid19_rango_edad.csv')
gdf = gpd.read_file('shapefiles_ccaa_espana.geojson')
```

Coordinates for mapping each CCAA

```
locations = {'Andalucía':[37.38,-5.97],
             'Aragón':[41.64,-0.88],
             'Asturias':[43.36,-5.85],
             'Balears':[39.57,2.65],
             'Canarias':[28.09,-15.41],
             'Cantabria':[43.46,-3.8],
             'Castilla La Mancha':[38.98,-3.92],
             'Castilla y León':[41.65,-4.77],
             'Cataluña':[41.39,2.17],
             'Ceuta':[35.89,-5.34],
             'C. Valenciana':[39.37,-0.8],
             'Extremadura':[39.71,-6.16],
             'Galicia':[43.12,-8.46],
             'Madrid':[40.49,-3.71],
             'Melilla':[35.29,-2.95],
             'Murcia':[38.00,-1.48],
             'Navarra':[42.66,-1.64]}
```

```

recovered=[42.00,-2.01],
'País Vasco':[43.23,-2.85],
'La Rioja':[42.27,-2.51]]}

```

```

def dateplot(x, y, **kwargs):
    ax = plt.gca()
    data = kwargs.pop("data")
    data.plot(x=x, y=y, ax=ax, grid=False, **kwargs)

```

```

infected['fecha'] = pd.to_datetime(infected['fecha'])
hospitalized['fecha'] = pd.to_datetime(hospitalized['fecha'])
uci['fecha'] = pd.to_datetime(uci['fecha'])
recovered['fecha'] = pd.to_datetime(recovered['fecha'])
death['fecha'] = pd.to_datetime(death['fecha'])

```

As the following variables follow an exponential line, the following plots will be displayed using logarithmic scale to see the lineal trend.

▼ Infected over time

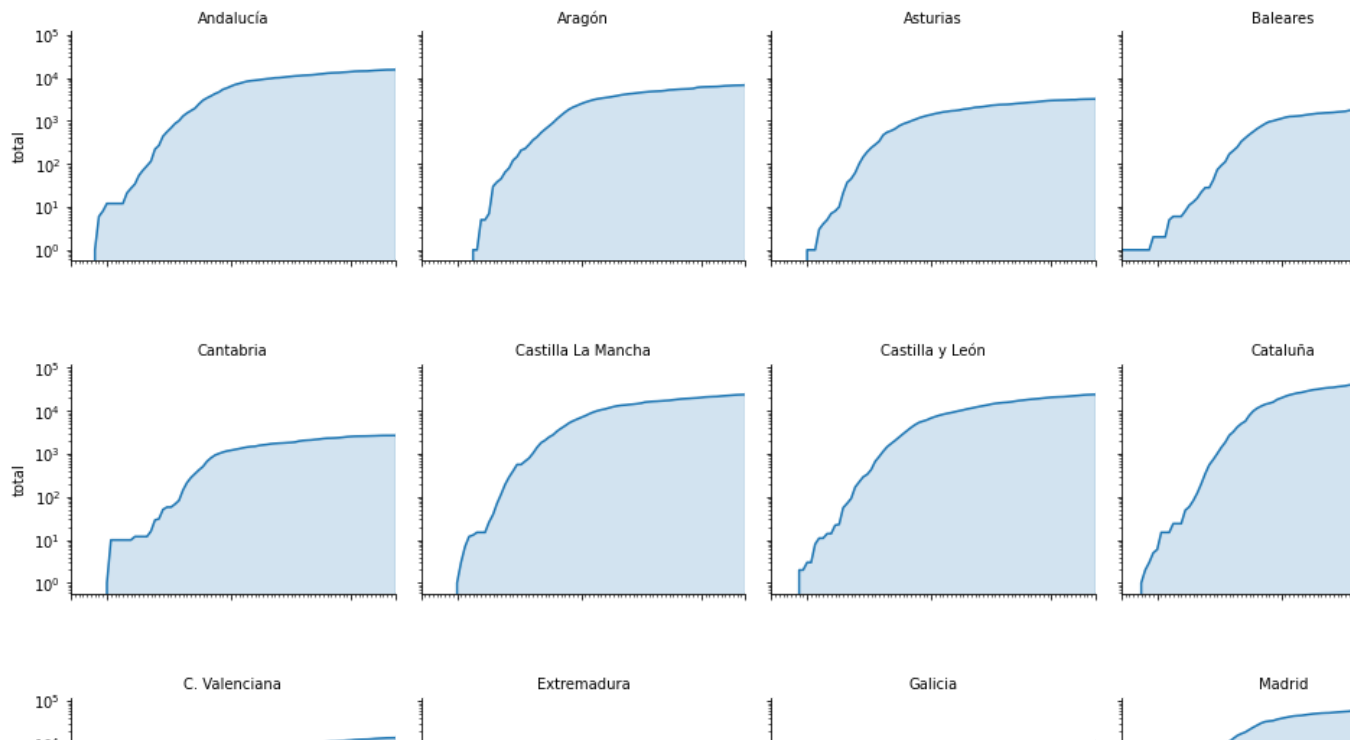
```

infected = infected[infected['CCAA']!= 'Total']
g = sns.FacetGrid(infected, col="CCAA", col_wrap=5, height=3.5)
g = g.map_dataframe(dateplot, "fecha", "total").set(yscale='log')
g = g.map(plt.fill_between, 'fecha', 'total', alpha=0.2).set_titles("{col_name} CCAA")
g = g.set_titles("{col_name}")
plt.subplots_adjust(top=0.92)
g = g.fig.suptitle('Evolution of total infected in CCAA (log scale)')

```



Evolution of total infected in CCAA (log scale)



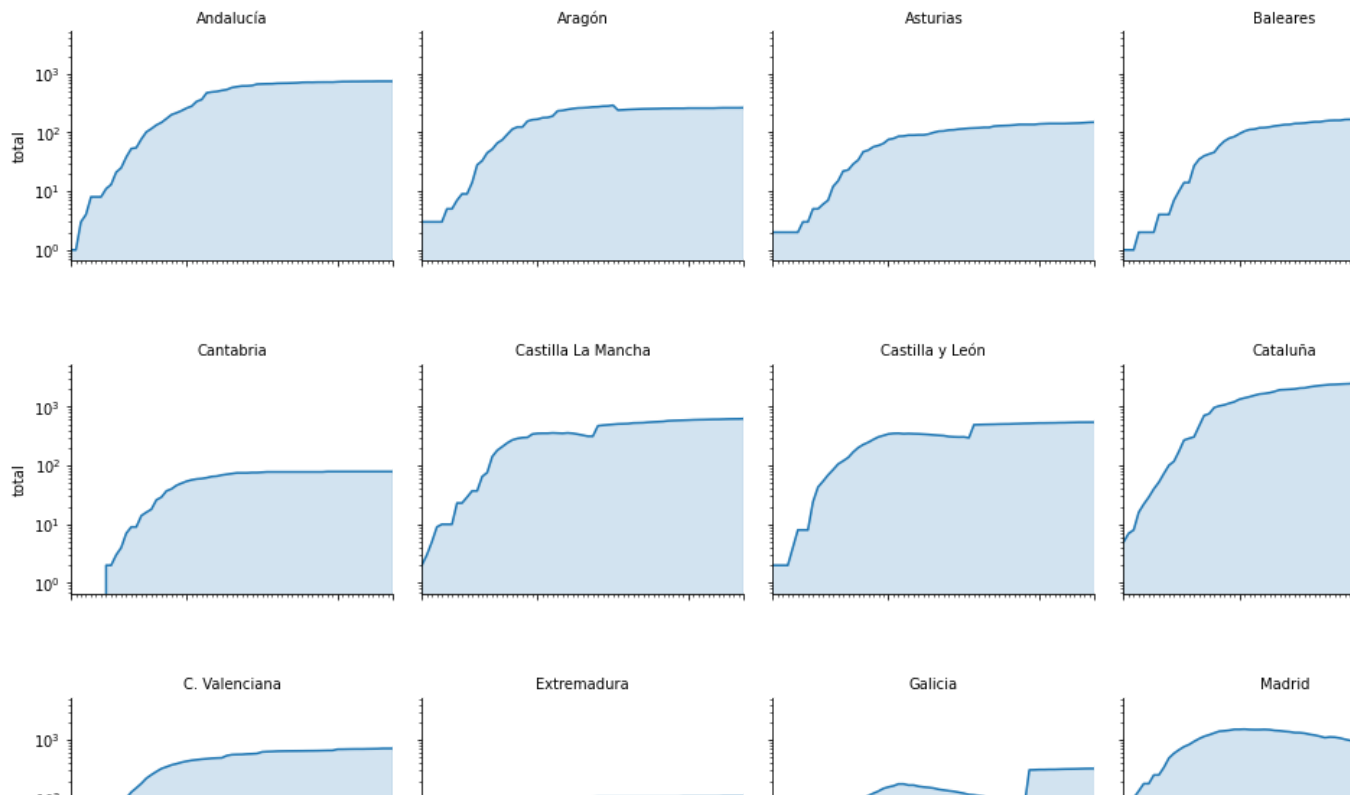
▼ ICU over time



```
uci = uci[uci['CCAA']!= 'Total']
g = sns.FacetGrid(uci, col="CCAA", col_wrap=5, height=3.5)
g = g.map_dataframe(dateplot, "fecha", "total").set(yscale='log')
g = g.map(plt.fill_between, 'fecha', 'total', alpha=0.2).set_titles("{col_name} CCAA")
g = g.set_titles("{col_name}")
plt.subplots_adjust(top=0.92)
g = g.fig.suptitle('Evolution of total UCI patients in CCAA (log scale)')
```



Evolution of total UCI patients in CCAA (log scale)



▼ Hospitalized over time



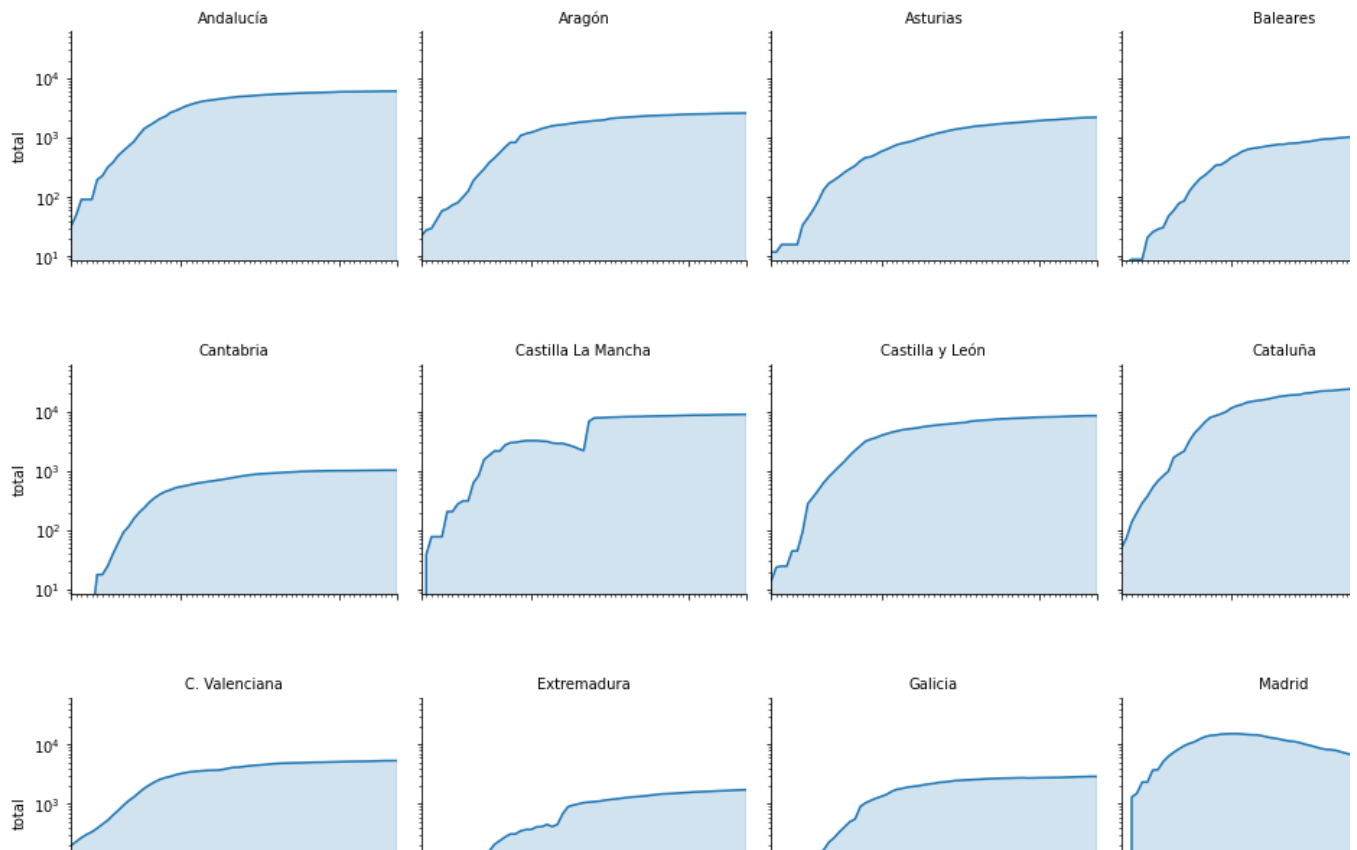
```

hospitalized = hospitalized[hospitalized['CCAA']!= 'Total']
g = sns.FacetGrid(hospitalized, col="CCAA", col_wrap=5, height=3.5)
g = g.map_dataframe(dateplot, "fecha", "total").set(yscale='log')
g = g.map(plt.fill_between, 'fecha', 'total', alpha=0.2).set_titles("{col_name} CCAA")
g = g.set_titles("{col_name}")
plt.subplots_adjust(top=0.92)
g = g.fig.suptitle('Evolution of total hospitalized in CCAA (Log Scale) ')

```



Evolution of total hospitalized in CCAA (Log Scale)



▼ Recovered over time

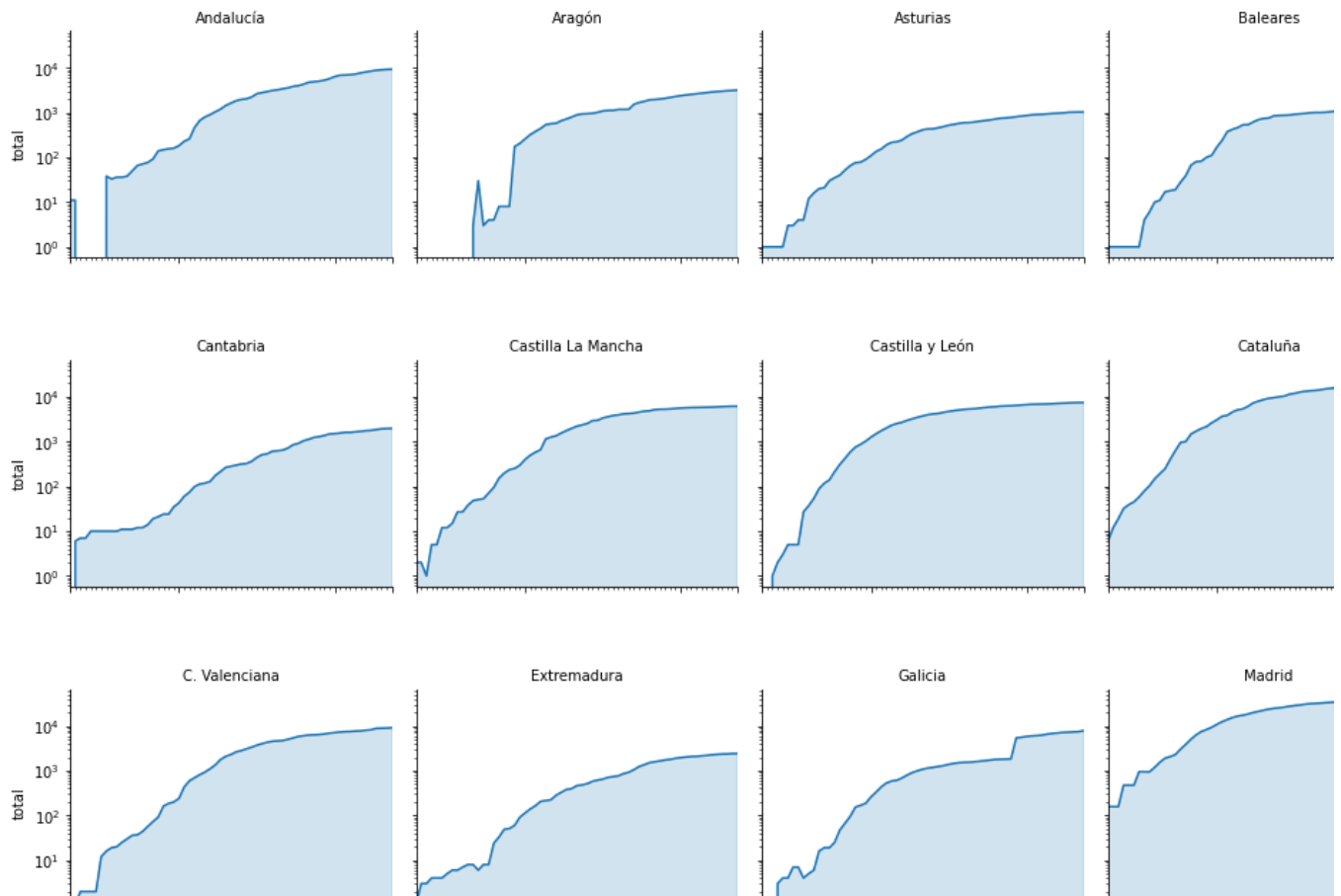
```

recovered = recovered[recovered['CCAA']!= 'Total']
g = sns.FacetGrid(recovered, col="CCAA", col_wrap=5, height=3.5)
g = g.map_dataframe(dateplot, "fecha", "total").set(yscale='log')
g = g.map(plt.fill_between, 'fecha', 'total', alpha=0.2).set_titles("{col_name} CCAA")
g = g.set_titles("{col_name}")
plt.subplots_adjust(top=0.92)
g = g.fig.suptitle('Evolution of total recovered in CCAA (Log Scale)')

```



Evolution of total recovered in CCAA (Log Scale)

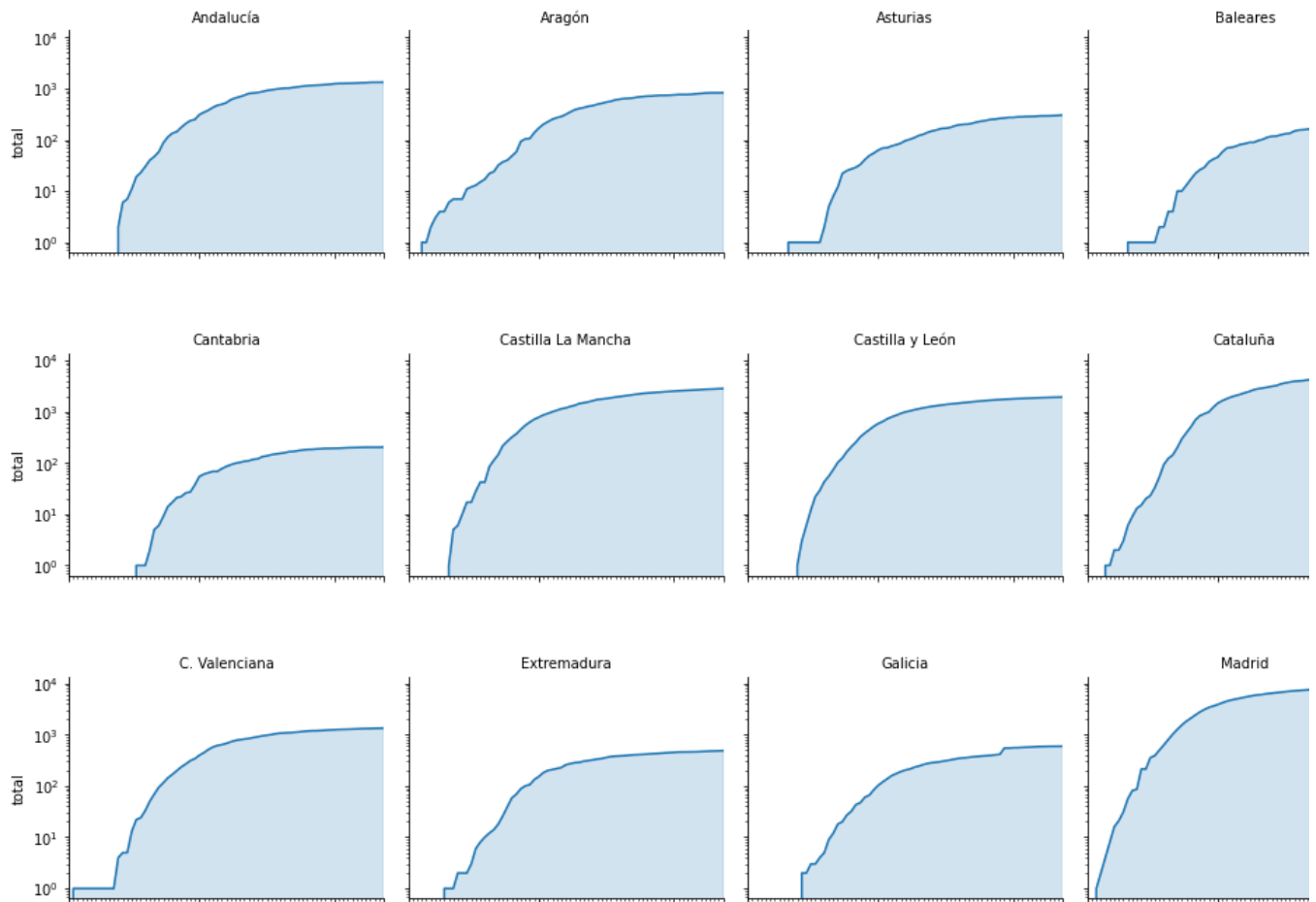


▼ Deaths over time

```
death = death[death['CCAA']!= 'Total']
g = sns.FacetGrid(death, col="CCAA", col_wrap=5, height=3.5)
g = g.map_dataframe(dateplot, "fecha", "total").set(yscale='log')
g = g.map(plt.fill_between, 'fecha', 'total', alpha=0.2).set_titles("{col_name} CCAA")
g = g.set_titles("{col_name}")
plt.subplots_adjust(top=0.92)
g = g.fig.suptitle('Evolution of total deaths in CCAA (log scale)')
```



Evolution of total deaths in CCAA (log scale)



▼ Patient Analysis



As a first approach, I will compare the effects of COVID-19 in each age group without taking into account

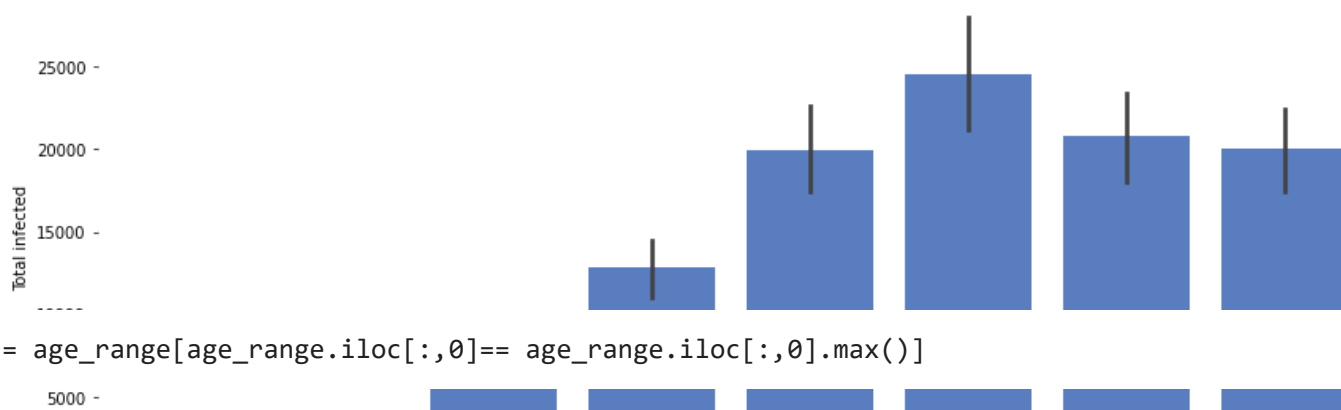


```
age_range= age_range[age_range['rango_edad']!='Total']
age_range= age_range[age_range['rango_edad']!='80 y +']
no_gender = age_range[age_range['sexo']=='ambos']
```

```
g = sns.catplot(x="rango_edad", y="casos_confirmados", hue="sexo", data=no_gender, kind="bar")
g.despine(left=True)
g.set_ylabels("Total infected")
```



```
<seaborn.axisgrid.FacetGrid at 0x7fa53d6e27b8>
```



```
last = age_range[age_range.iloc[:,0]== age_range.iloc[:,0].max()]
```

In case dataframe format is wrong:

```
for i in range(last['ingresos_uci'].shape[0]):
    if last.iloc[i,5] == 'i':
        last.iloc[i,5] = 0
```

```
last['ingresos_uci']= last['ingresos_uci'].astype(int)
```

In order to compare between different categories, we should normalize the data:

```
last['casos_confirmados'] = last['casos_confirmados'] / np.linalg.norm(last['casos_confirmados'])
last['hospitalizados'] = last['hospitalizados'] / np.linalg.norm(last['hospitalizados'])
last['ingresos_uci'] = last['ingresos_uci'] / np.linalg.norm(last['ingresos_uci'])
last['fallecidos'] = last['fallecidos'] / np.linalg.norm(last['fallecidos'])
```

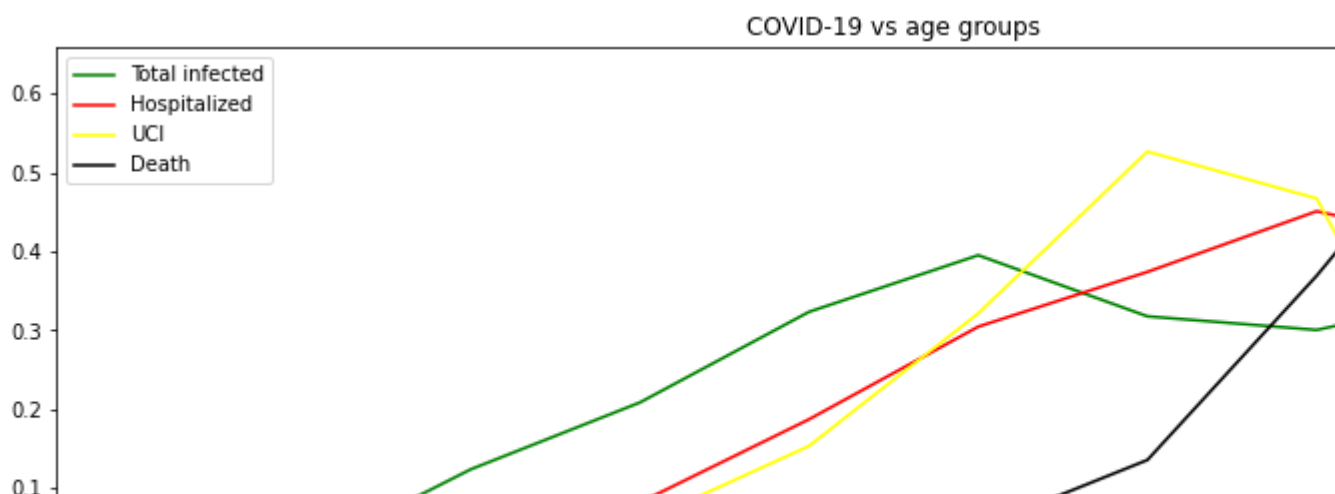
```
last_ambos = last[last['sexo']=='ambos']
last_gender = last[last['sexo']!='ambos']
```

▼ COVID-19 vs age groups

```
plt.figure(figsize=(15,5))
plt.plot(last_ambos['rango_edad'], last_ambos['casos_confirmados'],color = 'green',label='Total')
plt.plot(last_ambos['rango_edad'], last_ambos['hospitalizados'],color = 'red',label='Hospital')
plt.plot( last_ambos['rango_edad'], last_ambos['ingresos_uci'],color = 'yellow',label='UCI')
plt.plot( last_ambos['rango_edad'], last_ambos['fallecidos'],color = 'black',label='Death')
plt.title('COVID-19 vs age groups')
plt.legend()
```




<matplotlib.legend.Legend at 0x7fa53efb6ef0>

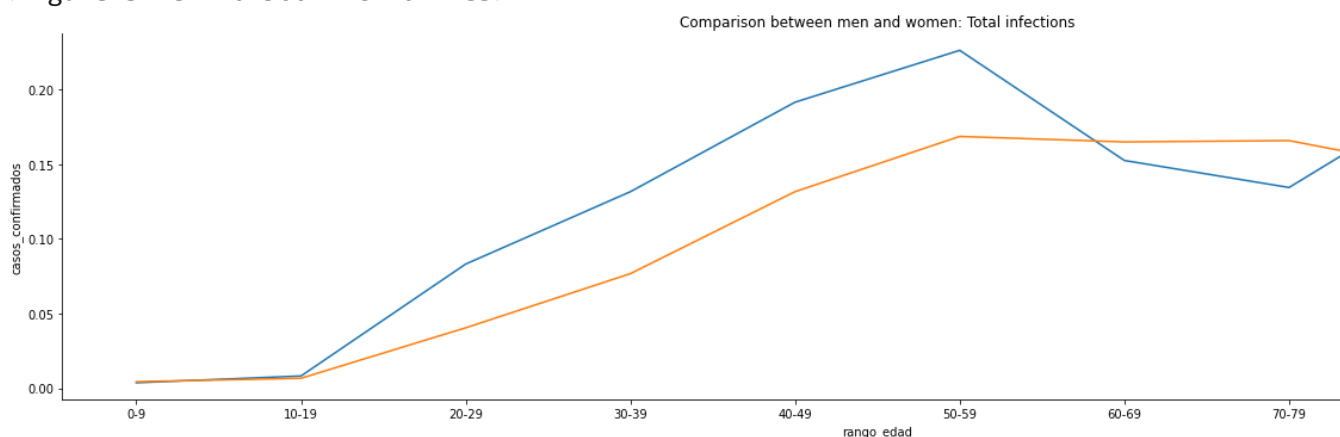


The results are as expected. virus affects mostly older people. We can see that the death's curve peaks at 70-79. A surprising result is the "Total infected" line as the 'peak' goes from 40 to 80 years which means the distribution of the population.

▼ COVID-19 vs Gender

```
plt.figure(figsize= (10,5))
sns.relplot(x='rango_edad',y ='casos_confirmados', hue = 'sexo',kind='line',data = last_gender)
plt.title('Comparison between men and women: Total infections')
```

 Text(0.5, 1.0, 'Comparison between men and women: Total infections')
<Figure size 720x360 with 0 Axes>

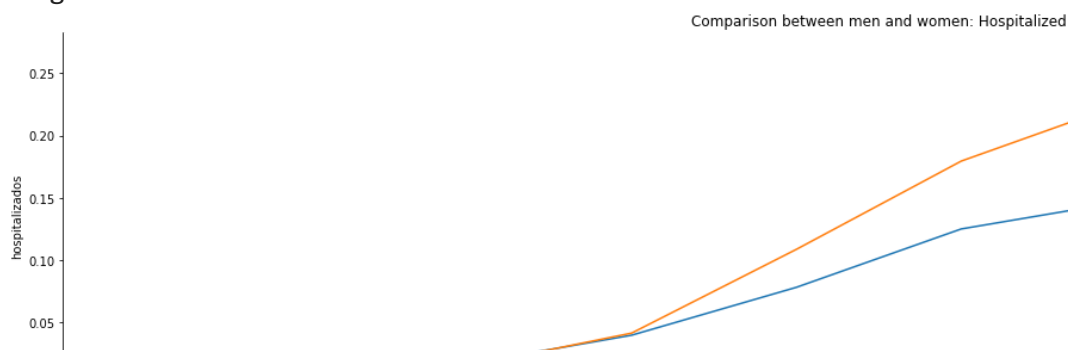


```
plt.figure(figsize= (10,5))
sns.relplot(x='rango_edad',y ='hospitalizados', hue = 'sexo',kind='line',data = last_gender)
plt.title('Comparison between men and women: Hospitalized')
```



Text(0.5, 1.0, 'Comparison between men and women: Hospitalized')

<Figure size 720x360 with 0 Axes>



```
plt.figure(figsize= (10,5))
```

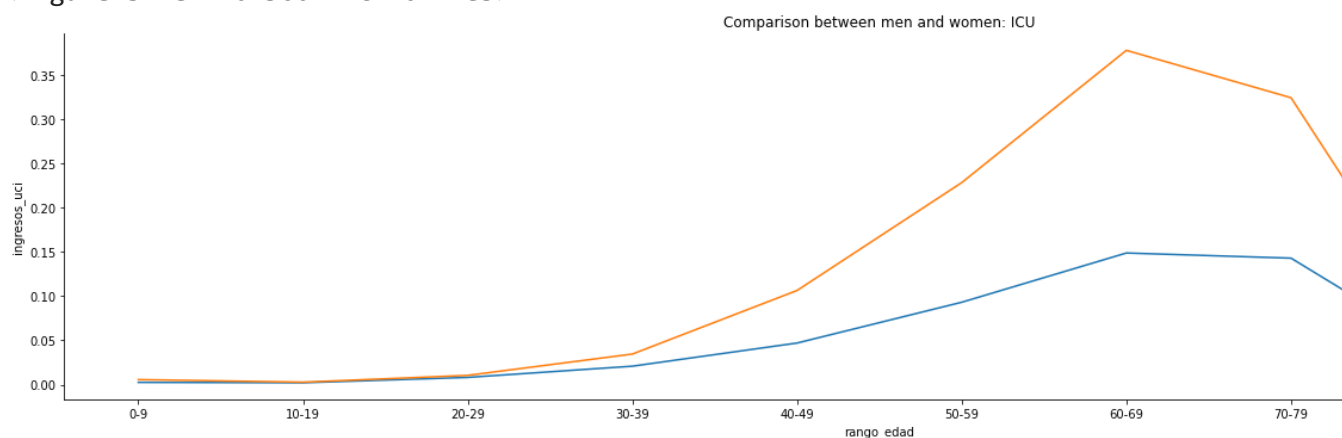
```
sns.relplot(x='rango_edad',y = 'ingresos_uci', hue = 'sexo',kind='line',data = last_gender,hei
```

```
plt.title('Comparison between men and women: ICU')
```



Text(0.5, 1.0, 'Comparison between men and women: ICU')

<Figure size 720x360 with 0 Axes>



```
plt.figure(figsize= (10,5))
```

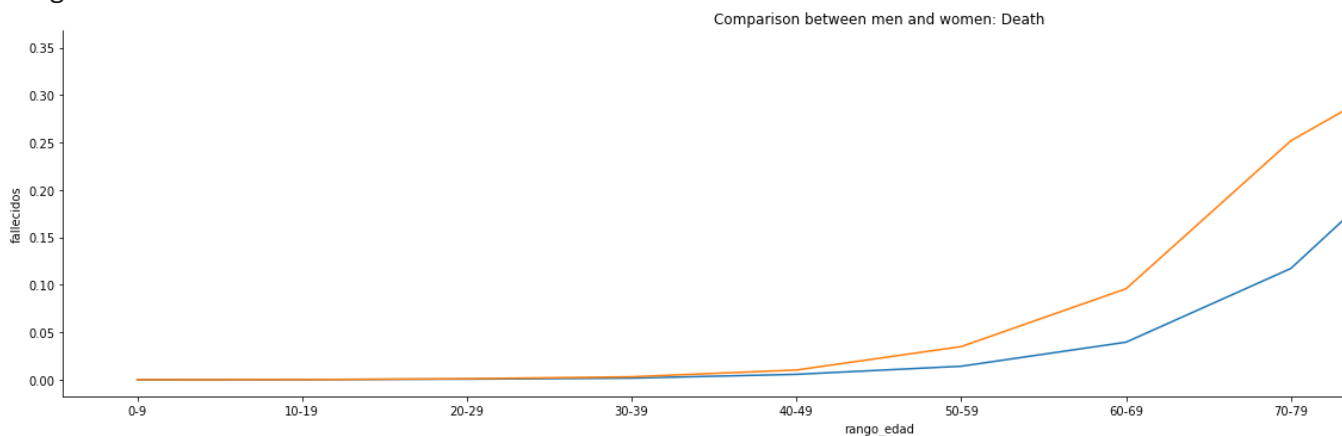
```
sns.relplot(x='rango_edad',y = 'fallecidos', hue = 'sexo',kind='line',data = last_gender,heigh
```

```
plt.title('Comparison between men and women: Death')
```



Text(0.5, 1.0, 'Comparison between men and women: Death')

<Figure size 720x360 with 0 Axes>



As we can see, in general the virus is more dangerous in men than woman although the total infected | 60yo. This can be easily explained with the previous chart as the age ranges in which women are mo less dangerous ages. One the other hand, the are more men than woman infected in the dangerous a