

# Cálculos, plots, group-by, y multi-indices

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# Interpolar Series de Tiempo

- dataframes de serie de tiempo
  - indices con `datetime`
  - posiblemente será muchos NaNs debido la hora de colecciona de datos
- interpolación para llenar valores de NaN values con otros valores
  - palabra de clave `method` para metodo de interpolación
  - `linear`, `time`, `pad`, etc.

```
# this is asreeve's custom module for reading levellogger files
2 from read_logger_files import to_dataframe
import pandas as pd
4
# create a data structure to hold individual data sets, and read them
  into dictionary
6 # to_dataframe returns a tuple contining the data set and header info
data_dict = {}
8 data_dict["shal"] = to_dataframe("examples/data/
  MSH_shallow_may2020Compensated.xle")[0]
data_dict["deep"] = to_dataframe("examples/data/
  MSH_deep_may2020Compensated.xle")[0]
10
# put the data into a dataframe with a multiindex
12 df = pd.DataFrame()
for depth in data_dict.keys():
14   data_dict[depth] = data_dict[depth].drop("block", axis=1)
   m_idx = pd.MultiIndex.from_product([[depth], data_dict[depth].
     columns])
16   data_dict[depth].columns = m_idx
   df = pd.concat((df, data_dict[depth]))
18 # change index stings to datetime objects and sort index
df.index = pd.to_datetime(df.index)
20 df = df.sort_index()
# interpolate the deep data onto the times shallow data was recorded
22 df.deep = df.deep.interpolate(method="time")
# drop rows with NaN values, save data
24 df = df.dropna()
df.to_hdf("examples/data/MSH_shal&deep_may2020.hdf", key="root")
26
# reampling aggregates data based on time
28 # to get daily mean values
df.resample('d').mean()
```

- aplicación permite funciones pesadas sobre subconjunto de dataframe
- tipo de ventana cambia el peso de datos con posición de muestro
- 'ventana' desliza por una columna de datos

```
import numpy as np
2 import pandas as pd
import matplotlib.pyplot as plt
4 import scipy.signal.windows as win

6 df = pd.read_hdf("examples/data/MSH_shal&deep_may2020.hdf")
# lets look at the multiindexed data,
8 df.plot(y=[("deep", "Lev"), ("shal", "Lev")])
# or use matplotlib directly
10 fig, sp = plt.subplots()
sp.plot(df.index, df.loc[:, ("shal", "Lev")], label="shallow")
12 sp.plot(df.index, df.loc[:, ("deep", "Lev")], label="deep")
sp.legend()
14 sp.set_ylabel("Water Pressure (m of H2O)")
plt.close()

16
# make a rolling window, calculate mean
18 a_win = df.loc[:, ("shal", "Lev")].rolling(9, center=True)
a_roll = a_win.agg("mean")
20
# can do other window types defined in scipy.signal.windows library
22 b_win = df.loc[:, ("shal", "Lev")].rolling(9, center=True, win_type="triang")
b_roll = b_win.agg("mean")
24

df = df.assign(box=a_roll)
26 df = df.assign(tri=b_roll)

28 fig, sp = plt.subplots()
sp.plot(df.index, df.loc[:, "box"], label="box")
30 sp.plot(df.index, df.loc[:, "tri"], label="tri")
sp.plot(df.index, df.loc[:, ("shal", "Lev")], label="shallow")
32 sp.legend()
sp.set_ylabel("Water Pressure (m of H2O)")
```

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```
# continued from previous script
2 # rolling function using two columns (correlation)
  A=df.loc[:,('deep','Lev')].rolling(10,center=True).corr(df.loc[:,('
    shal','Lev')])
4 df=df.assign(corr=A)

6 # using your own function, a rolling IQR
  def iqr(x):
8     value=np.quantile(x,.75)-np.quantile(x,.25)
    return value
10
  A=df.loc[:,('deep','Lev')].rolling(10,center=True).agg(iqr)
12 df=df.assign(iqr=A)

14 fig, sp = plt.subplots(nrows=2,ncols=1, sharex=True)
  sp[0].plot(df.index, df.loc[:, "box"], label="box")
16 sp[0].plot(df.index, df.loc[:, "tri"], label="tri")
  sp[0].plot(df.index, df.loc[:, ("shal", "Lev")], label="shallow")
18 sp[0].plot(df.index, df.loc[:, ("deep", "Lev")], label="deep")
  sp[0].legend()
20 sp[0].set_ylabel("Water Pressure (m of H2O)")

22 sp[1].plot(df.index, df.loc[:, 'iqr'],color='black')
  sp[1].set_ylabel('interquartile range', color='black')
24
  sp1tw=sp[1].twinx()
26 sp1tw.plot(df.index, df.loc[:, 'corr'],color='red')
  sp1tw.set_ylabel('correlation',color='red')
28 sp1tw.yaxis.set_tick_params(labelcolor='red')
  plt.show()
```

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# Ploteas elegantes con Seaborn

- seaborn: hace diferentes ploteas estadísticas
- aparece mejor? que matplotlib
- hecho para funcionar bien con pandas
- ploteas complicadas, hecho sencillas!

```
import pandas as pd
2 import matplotlib.pyplot as plt
import seaborn as sns
4 # read in data and get into usable format , as done before
df = pd.read_table("../examples/USGS01037000.tab", comment="#")
6 col_dict = {}
for col in df.columns:
8     if ("_00065" in col) and ("_cd" not in col):
        col_dict[col] = "gage_ht_ft"
10     elif ("_00060" in col) and ("_cd" not in col):
        col_dict[col] = "Q_cfs"
12 df = df.rename(col_dict, axis=1)
df = df.loc[:, ["datetime", "Q_cfs", "gage_ht_ft"]]
14 df.loc[:, "datetime"] = pd.to_datetime(df.loc[:, "datetime"])
df.loc[:, "gage_ht_ft"] = pd.to_numeric(df.loc[:, "gage_ht_ft"], errors="coerce")
16 df.loc[:, "Q_cfs"] = pd.to_numeric(df.loc[:, "Q_cfs"], errors="coerce")
df = df.assign(month=df.loc[:, "datetime"].dt.month)
18
sns.set_theme(style="darkgrid")
20 # making a nice looking regression plot
fig, sps = plt.subplots(ncols=1, nrows=2)
22 # can mix seaborn and matplotlib plots
sps[0].xaxis.update_units(df.datetime)
24 sps[0].semilogy(df.datetime.values, df.Q_cfs.values)
# seaborn regresion plot, display minor gridlines
26 sns.regplot(x=df.gage_ht_ft, y=df.Q_cfs, ax=sps[1], order=1, scatter_kws={"color": "black"}, line_kws={"color": "red"})
sps[0].grid(which="both", axis="y")
28 plt.show()
```

# Dataframes con Mult-indices

- niveles diferentes con nombres de indices y columnas
- maneja mas que 2-D datos
- configurando unos indices de 'multi-level' indices y columna es diferente
  - sobrescribe el encabezamiento de columna
  - para indice, lista las columnas que quiere para 'multi-index'

```
import pandas as pd
2 import seaborn as sb
# increase number of visible rows
4 pd.options.display.min_rows = 30

6 # get one of the seaborn data sets, need internet connection
# list available data sets with 'sb.get_dataset_names()'
8 df = sb.load_dataset("penguins")
# drop rows with NaN values, get a list of unique island names
10 df = df.dropna(axis=0, how="any")
islands = df.island.unique()
12 # overwrite original index with multiindex created from columns of
    data
df2 = df.set_index(["island", "sex", "species"])
14 # extract data for each island, create multiindex for each island, and
    concat
island_dict = {}
16 df3 = pd.DataFrame()
for isle in islands:
18     island_dict[isle] = df2.loc(axis=0)[isle]
    # multi-index has to have the same number of columns as dataframe
    assigned to
20 # there are several ways in pandas to create the multiindex, this
    is just one
m_idx = pd.MultiIndex.from_product([[isle], island_dict[isle].
    columns])
22 # overwrite original column with multilevel columns (multiindex)
island_dict[isle].columns = m_idx
24 df3 = pd.concat((df3, island_dict[isle].iloc[:, 0:2]))
df3 = df3.sort_index()
```

# Agrupar y Agregar los Datos

- agrega los valores (sum, mean, variance, etc.)
- metodo de groupby
  - agrupa valores juntos

```
import pandas as pd
2 import numpy as np

4 # read in data and get into usable format
df = pd.read_table("../example/USGS01037000.tab", comment="#")
6
# rename columns
8 col_dict = {}
for col in df.columns:
10     if ("_00065" in col) and ("_cd" not in col):
        col_dict[col] = "gage_ht_ft"
12     elif ("_00060" in col) and ("_cd" not in col):
        col_dict[col] = "Q_cfs"
14 df = df.rename(col_dict, axis=1)

16 df = df.loc[:, ["datetime", "Q_cfs", "gage_ht_ft"]]
df.loc[:, "datetime"] = pd.to_datetime(df.loc[:, "datetime"])
18 df.loc[:, "gage_ht_ft"] = pd.to_numeric(df.loc[:, "gage_ht_ft"],
        errors='coerce')
df.loc[:, "Q_cfs"] = pd.to_numeric(df.loc[:, "Q_cfs"], errors='
        coerce')

20
# calculate aggregate values for columns
22 mean = df.mean()
median = df.median()
24 var = df.var()
# df.describe() for short table of summary stats
```

# Más Agrupar y Agregar los Datos

- usa `assign` para hacer columnas nuevas 'para agrupar'
- puede encadenar metodos `groupby` y `aggregate`
- puede hacer funciones personalizadas para agregar

```
##### CONTINUED FROM PREVIOUS CODE BLOCK #####
2 # Grouping data
# calculate monthly statistics:
4 # make column of months
df = df.assign(month=df.loc[:, "datetime"].dt.month)
6 # group by month
# splitting data, applying function, combinging into dataframe
8 g_month = df.groupby("month")
month_mean = g_month.agg("mean")
10 # or chain these together
# monthly_mean = df.groupby('month').agg('mean')
12
# calculate weekly stats
14 # careful with isocalander, all weeks start on monday,
# Dec 31 could be in the first week of the next year
16 df = df.assign(week=df.loc[:, "datetime"].dt.isocalendar()["week"])
week_stats = df.groupby("week")["Q_cfs", "gage_ht_ft"].agg([min, np.
    median, max])
18
# making a custom function
20 def my_slope(series):
    slope, inter = np.polyfit(series.index, series.values, 1)
22     return slope
24 week_stats2 = df.groupby("week")["Q_cfs", "gage_ht_ft"].agg(my_slope)
```

# Replantear, Desremar y Rebanar

- `unstack`: mueve una porción del índices al columnas
- `stack`: mueve una porción de columnas al índices
- usa `level` palabra clave para indicar niveles para mover
  - nivel cero es más externa de multi-index

```
import pandas as pd
2 # increase number of visible rows
pd.options.display.min_rows = 30
4 # get one of the seaborn data sets
df = sb.load_dataset("penguins")
6 # drop rows with NaN values, change to multiindex
df = df.dropna(axis=0, how="any")
8 df1 = df.set_index(["species", "sex", "island"])
# multiindex dataframes perform better if sorted
10 df1 = df1.sort_index()
# get into a series format
12 df2 = df1.stack()
# Can't unstack df2 because has duplicate values, can't set columns
to same value
14 # to get around this, include original index in reindexed dataframe
df3 = df.set_index([df.index, "species", "sex", "island"])
16 df3 = df3.sort_index()
# can control level(s) unstacked and moved to column heading with
keyword: level=2 or level=[1,2]
18 df4 = df3.unstack()
df5 = df4.stack()
20 ## slicing the dataframe
A = df3.loc[(0, "Adelie", "Male", "Torgersen"):(19, "Adelie", "Male",
" Torgersen"),
"bill_length_mm".]
22 # if duplicate values in dataframe, can specify axis to get all
duplicated values
24 B = df2.loc["Adelie"]
C = df2.loc(axis=0)["Adelie", "Male"]
26 # also need to specify axis to be able slice on 2nd variable
D = df1.loc(axis=0)[:, "Male"]
28 E = df1.loc(axis=0)[:, "Male"]["body_mass_g"]
```

# Swaplevel y Reorderlevel

- para cambiar la organización de niveles de multi-index
  - `swaplevel` intercambia dos niveles
  - `reorderlevel` intercambia más que dos niveles
    - lista niveles en orden que quiere
    - `axis` palabra clave para indicar índices de filas o columnas
    - puede ordenar índices de columna con `sort_index(axis=1)`

```
import pandas as pd
import seaborn as sb

# increase number of visible rows
pd.options.display.min_rows = 30

# get one of the seaborn data sets
df1 = pd.read_hdf('./data/penguins.hdf')

# drop rows with NaN values, change to multiindex
df2 = df1.set_index([df1.index, "species", "sex", "island"])

df2 = df2.sort_index()
# can control level(s) that is unstacked and moved to column heading
# with the keyword
# level level=2 or level=[1,2]
df2 = df2.unstack()

# commands to change ordering in indices
df2.swaplevel(0,2)
df2.swaplevel(axis=1)
df2.swaplevel(axis=1).sort_index(axis=1)

#note extra parentheses
df2.reorder_levels((1,0,2))
```

# Xs, Pivot y Pivot\_table

- sección transversal (xs): subconjunto de multi-index
  - solo función con multi-index
- pivot: remodelar dataframe
  - requiere valores únicos
  - todos los valores vuelven
- pivot\_table:
  - remodelar dataframe
  - agrega valores duplicados

```
import pandas as pd
2 import seaborn as sb
import numpy as np

4
df0 = sb.load_dataset("penguins")
6 df0 = df0.dropna()
df1 = df0.set_index([df0.index, "sex", "species"])
8 df1 = df1.sort_index()
# xs only works on a multiindex, selects subset of values
10 df1.xs("Gentoo", level=2)
df1.xs(["Female", "Gentoo"], level=[1, 2])
12
# reshape df, pivot requires no duplicate values with no aggregation
14 # pivot_table will aggregate duplicate data (defaults to mean)
df1.pivot_table(values=["bill_length_mm", "bill_depth_mm"], index=["sex"])
16 df1.pivot_table(
    values=["bill_length_mm", "bill_depth_mm"], index=["sex", "island"],
    aggfunc=np.std
18 )
# dummy data to show pivot method
20 df2 = pd.DataFrame(
    {
22     "Na": np.random.uniform(0, 1, 10),
    "Ca": np.random.uniform(0, 1, 10),
24     "K": np.random.uniform(0, 1, 10),
    "Mg": np.random.uniform(0, 1, 10),
26     "loc": [1, 2, 3, 4, 5] * 2,
    "depth": [0, 0, 0, 0, 0, 1, 1, 1, 1, 1],
28     }
    )
30 df2.pivot(index="loc", columns="depth", values=["K", "Na"])
```