```
In [1]:
        Author Nicola Lombardi, SW HW Engineer Computer and Telecommunications Engineer
        & Senior Python Dev.
        _____
        Telecomunicazioni / Elettronica / Informatica
        Responsabile Cost&Qualità
        Ex Ingegnere Hardware Baseband
        Web: Curriculum pubblico
        contributi scientifici: Ricerca
        ______
        -> Collab: Andrea Casula, Paolo Pintus
        Professor Paolo Pintus (Assistant Professor dipartimento di Fisica)
        Web - https://web.unica.it/unica/page/it/paolo_pintus
        Relazione con il professore: Relatore di tesi del 2022
              - Ricercatore a tempo determinato
        Area scientifico disciplinare - Scienze fisiche
        Settore scientifico disciplinare FIS/03 FISICA DELLA MATERIA
        Email paolo.pintus@unica.it
        Professor Andrea Casula (Associate Professor in Sistemi Wireless, Laurea Magistrale
        "Internet Engineering")
        Web - https://web.unica.it/unica/it/ateneo_s07_ss01.page?contentId=SHD30253
        Relazione con il professore: Relatore di tesi del 2022
        Ruolo - Professore associato
        Area scientifico disciplinare - Ingegneria industriale e dell'informazione
        Settore scientifico disciplinare - ING-INF/02 CAMPI ELETTROMAGNETICI
        DATE: February 2025 NAME OF THE PROJECT: REAL TIME BLE TRACKER SYSTEM
        .....
       "\nAuthor Nicola Lombardi, SW HW Engineer Computer and Telecommunications Engineer \n
Out[1]:
       & Senior Python Dev. \n\n======\n
       Telecomunicazioni / Elettronica / Informatica\nResponsabile Cost&Qualità \nEx Ingegn
       ere Hardware Baseband \nWeb: Curriculum pubblico\ncontributi scientifici: Ricerca\n==
       =======\n\n-----\n\n-----
       -----\n-> Collab: Andrea Casula, Paolo Pintus\n-----
       -----\n\n\nProfessor Paolo Pintus (Assistant Professor dipartimento di Fisica) \nW
       eb - https://web.unica.it/unica/page/it/paolo pintus \nRelazione con il professore: R
       elatore di tesi del 2022 \nRuolo - Ricercatore a tempo determinato \nArea scientif
       ico disciplinare - Scienze fisiche \nSettore scientifico disciplinare FIS/03 FISIC
       A DELLA MATERIA \nEmail paolo.pintus@unica.it\n\nProfessor Andrea Casula (Associa
       te Professor in Sistemi Wireless, Laurea Magistrale \n"Internet Engineering") \nWeb
           https://web.unica.it/unica/it/ateneo_s07_ss01.page?contentId=SHD30253 \nRelazione
```

con il professore: Relatore di tesi del 2022 \nRuolo - Professore associato \nArea s cientifico disciplinare - Ingegneria industriale e dell'informazione \nSettore scie

- ING-INF/02 CAMPI ELETTROMAGNETICI \n\n\nDATE: February 20

In [2]:

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from SALib.analyze import rbd_fast

25 NAME OF THE PROJECT: REAL TIME BLE TRACKER SYSTEM\n\n"

ntifico disciplinare

```
# this is where you define your own model, procedure, experiment...
from SALib.test_functions import Ishigami
import numpy as np
```

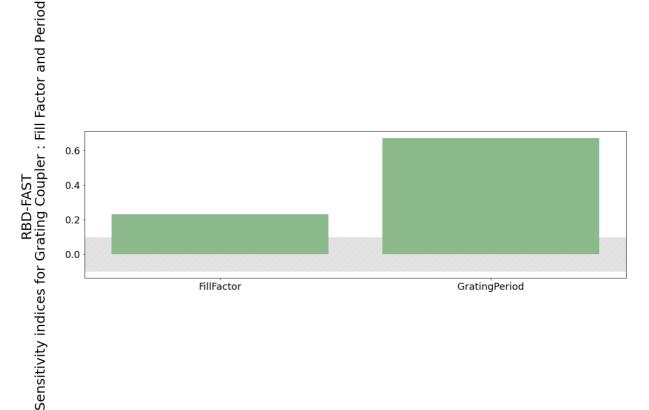
```
In [3]:
         # this is where you define your own model, procedure, experiment...
         from SALib.test_functions import Ishigami
         import numpy as np
         def run model(x1, x2):
            COPY HERE YOUR OWN CODE
            the function takes as input 1 sample
            returns 1 or more output
            lmbd_0 = 1.55 * (np.math.pow(10,-6)) # central wavelength is 1550 nanometers
            n1 = 2.848 # effective index of the grating teeth
            n2 = 2.534 # effective index of the grating slots
            # As an example, we'll look at the famous Ishigami function
            # (A and B are specific parameters for the Ishigami function)
            A = 7
            B = 0.1
            y = np.arcsin(((n1*x1+n2-x1*n2)*x2- lmbd_0)/x2)
            y_{deg} = 180/np.pi * y
            # ====== TO HERE and replace with your own piece of code
            return y_deg
         def conv_study(n, Y, X):
            # take n samples among the num_samples, without replacement
            subset = np.random.choice(num samples, size=n, replace=False)
            return rbd_fast.analyze(problem=problem,
                                    Y=Y[subset],
                                    X=X[subset])['S1']
         def bootstrap(problem, Y, X):
            Calculate confidence intervals of rbd-fast indices
            1000 draws
            returns 95% confidence intervals of the 1000 indices
            problem : dictionnary as SALib uses it
            X : SA input(s)
            Y : SA output(s)
            all_indices = []
            for i in range(1000):
                X new = np.zeros(X.shape)
                Y new = np.zeros(Y.shape).flatten()
```

```
# draw with replacement
                 tirage_indices = np.random.randint(0, high=Y.shape[0], size = Y.shape[0])
                 for j, index in enumerate(tirage_indices):
                     X_{\text{new}}[j,:] = X[\text{index},:]
                     Y_{new[j]} = Y[index]
                 all_indices.append(rbd_fast.analyze(problem=problem, Y=Y_new, X=X_new)['S1']
             means = np.array([i.mean() for i in np.array(all_indices).T])
             stds = np.array([i.std() for i in np.array(all_indices).T])
             return np.array([means - 2 * stds, means + 2 * stds])
In [4]:
         # STATE THE PROBLEM DICTIONNARY
         # what will be varying (=inputs)? in what bounds?
         FF_min = 0.2
         FF_max = 0.7
         g_period_power = (np.math.pow(10,-9)) # structure grating period is 660 nanometers
         g_period_min = 620 * g_period_power # 620 nm
         g_period_max = 690 * g_period_power # 690 nm
         problem = {
             'num_vars': 2,
             'names': ['FillFactor', 'GratingPeriod'],
             'bounds': [[FF_min,FF_max],
                        [g_period_min, g_period_max]]
         # say we want to draw 150 samples
         num samples = 150
         # we draw a Latin Hypercube Sampling (LHS) that is fitted for an RBD FAST analysis
         # (other sampling metods available in the library though)
         from SALib.sample.latin import sample
         all_samples = sample(problem, num_samples)
         # run your model, procedure, experiment for each sample of your sampling
         # unpack all samples into 3 vectors x1, x2, x3
         x1, x2 = all samples.T
In [5]:
         # run the model, all samples at the same time
         ishigami results = run model(x1, x2)
         # Let us look at the analyze method
         rbd_fast.analyze(problem=problem, Y=ishigami_results, X=all_samples)
         # storing the first order indices of the analyze method
         si1 = rbd_fast.analyze(problem=problem, Y=ishigami_results, X=all_samples)['S1']
         # make nice plots with the indices (looks good on your presentations)
         # do not use the plotting tools of SALib, they are made for the method of Morris ...
```

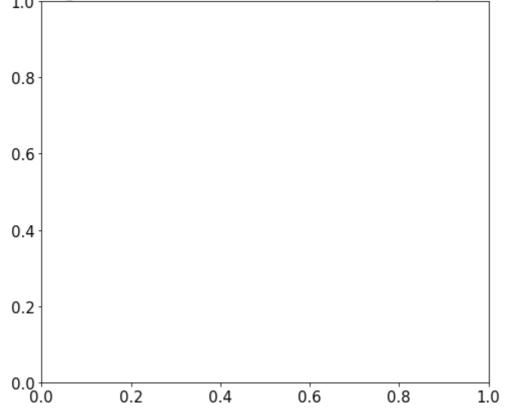
```
fig, ax = plt.subplots()
fig.set_size_inches(18, 5)
ax.tick_params(labelsize=18)
```

```
# ===== X-AXIS =====
ax.set_xticks(np.arange(problem['num_vars']))
ax.set_xticklabels(problem['names'])
ax.set_xlim(xmin=-0.5, xmax=problem['num_vars'] - 0.5)
# ===== Y-AXIS =====
ax.set_ylabel('RBD-FAST\nSensitivity indices for Grating Coupler : Fill Factor and P
# ==== BARS REPRESENTING THE SENSITIVITY INDICES =====
ax.bar(np.arange(problem['num_vars']), si1,
       color='DarkSeaGreen');
#in striped grey : not significant indices
ax.fill_between(x=[-0.5, 5.5], y1=-0.1, y2=0.1, color='grey', alpha=0.2, hatch='//',
# take a closer look to undestand interactions (looks even better on your presentati
# this part can be done without analyzing with SALib, just with the ouput from the s
fig, ax = plt.subplots()
fig.set_size_inches(8, 7)
ax.tick params(labelsize=15)
ax.set_title('Output of the model studied (Bragg condition)\n'
             'according to the value of the 2 most influent parameters',
             fontsize=20)
```

Out[6]: Text(0.5, 1.0, 'Output of the model studied (Bragg condition)\naccording to the value of the 2 most influent parameters')



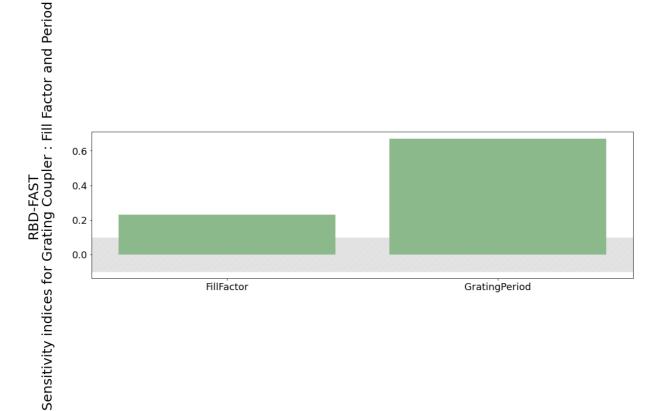
Output of the model studied (Bragg condition) according to the value of the 2 most influent parameters

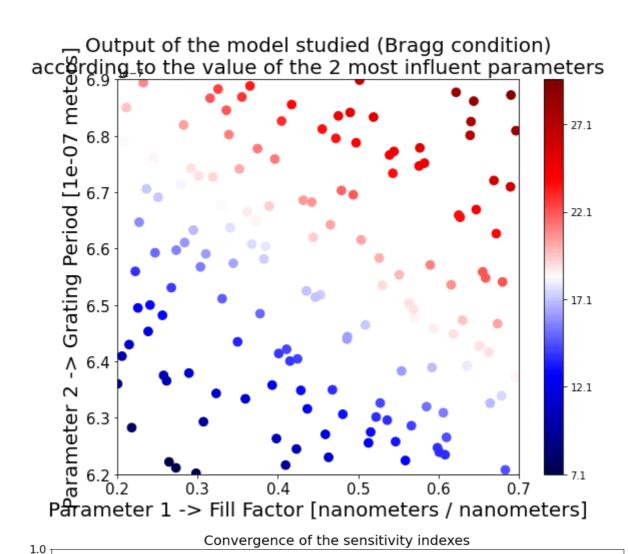


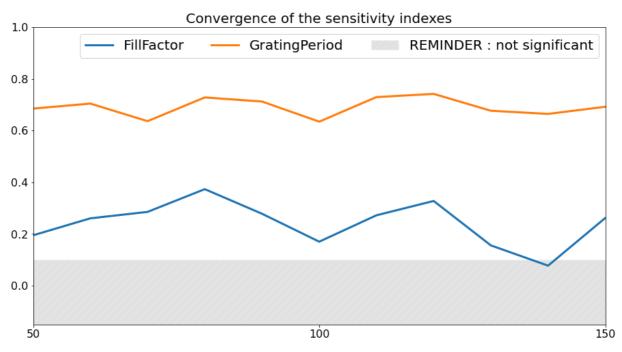
```
In [7]:
         fig, ax = plt.subplots()
         fig.set_size_inches(18, 5)
         ax.tick_params(labelsize=18)
         # ===== X-AXIS =====
         ax.set_xticks(np.arange(problem['num_vars']))
         ax.set_xticklabels(problem['names'])
         ax.set_xlim(xmin=-0.5, xmax=problem['num_vars'] - 0.5)
         # ===== Y-AXIS =====
         ax.set_ylabel('RBD-FAST\nSensitivity indices for Grating Coupler : Fill Factor and P
         # ==== BARS REPRESENTING THE SENSITIVITY INDICES =====
         ax.bar(np.arange(problem['num_vars']), si1,
                color='DarkSeaGreen');
         #in striped grey : not significant indices
         ax.fill between(x=[-0.5, 5.5], y1=-0.1, y2=0.1, color='grey', alpha=0.2, hatch='//',
         # take a closer look to undestand interactions (looks even better on your presentati
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         fig.set_size_inches(8, 7)
         ax.tick params(labelsize=15)
         ax.set_title('Output of the model studied (Bragg condition)\n'
                       'according to the value of the 2 most influent parameters',
                      fontsize=20)
```

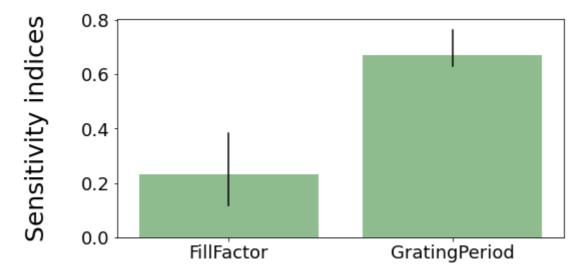
```
# ===== SCATTER =====
size = np.ones(num_samples) * 75
sc = ax.scatter(x1, x2,
                c=ishigami_results,
                s=size,
                vmin=ishigami_results.min(),
                vmax=ishigami_results.max(),
                cmap='seismic',
                edgecolor=None)
# ===== X-AXIS =====
ax.set_xlim(xmin=problem['bounds'][0][0], xmax=problem['bounds'][0][1])
ax.set xlabel('Parameter 1 -> Fill Factor [nanometers / nanometers]', fontsize=20)
# ===== Y-AXIS =====
ax.set_ylim(ymin=problem['bounds'][1][0], ymax=problem['bounds'][1][1])
ax.set_ylabel('Parameter 2 -> Grating Period [1e-07 meters]', fontsize=20)
# ===== COLORBAR =====
ticks = np.arange(ishigami_results.min(), ishigami_results.max(), 5)
cb = plt.colorbar(sc, ticks=ticks);
cb.ax.set_yticklabels([str(round(i,1)) for i in ticks])
fig.tight_layout()
all_indices = np.array([conv_study(n=n, Y=ishigami_results, X=all_samples)
                        for n in np.arange(50, num_samples + 1, 5)])
# convergence check
fig, ax = plt.subplots()
fig.set_size_inches(15,8)
ax.set title('Convergence of the sensitivity indexes', fontsize=20)
ax.tick_params(labelsize=16)
ax.set_xlim(xmin=0, xmax=(num_samples - 50)//10)
ax.set_xticks(np.arange(0,(num_samples - 50)//10 +1,5))
ax.set_xticklabels([str(i) for i in range(50,num_samples+1,50)])
ax.set ylim(ymin=-0.15, ymax=1)
for p,param in enumerate(problem['names']):
    ax.plot(all_indices[:,p], linewidth=3, label=param)
ax.fill_between(x=[0,(num_samples - 50)//10], y1=-0.15, y2=0.1, color='grey', alpha=
              label='REMINDER : not significant')
ax.legend(fontsize=20, ncol=4)
# Get bootstrap confidence intervals for each index
bootstrap_conf = bootstrap(problem=problem, Y=ishigami_results, X=all_samples)
```

```
# make nice plots with the indices (looks good on your presentations)
# do not use the plotting tools of SALib, they are made for the method of Morris ...
fig, ax = plt.subplots()
fig.set_size_inches(8,4)
ax.tick_params(labelsize=18)
# ===== X-AXIS =====
ax.set_xticks(np.arange(problem['num_vars']))
ax.set_xticklabels(problem['names'])
ax.set_xlim(xmin=-0.5, xmax=problem['num_vars'] - 0.5)
# ===== Y-AXIS =====
ax.set_ylabel('Sensitivity indices\n', fontsize=25)
# ==== BARS REPRESENTING THE SENSITIVITY INDICES =====
ax.bar(np.arange(problem['num_vars']), si1,
         color='DarkSeaGreen');
# ==== LINES REPRESENTING THE BOOTSTRAP "CONFIDENCE INTERVALS" =====
for j in range(problem['num_vars']):
    ax.plot([j, j], [bootstrap_conf[0, j], bootstrap_conf[1, j]],
\#ax.fill\_between(x=[-0.5, 5.5], y1=-0.1, y2=0.1, color='grey', alpha=0.2, hatch='//'
```









```
import numpy as np
def run_model(x1, x2):
   COPY HERE YOUR OWN CODE
   the function takes as input 1 sample
   returns 1 or more output
   \#lmbd_0 = 1.55 * (np.math.pow(10,-6)) \# central wavelength is 1550 nanometers
   # Lmbd_0 is the new x1
   FF=0.6
   n1 = 2.848 # effective index of the grating teeth
   n2 = 2.534 # effective index of the grating slots
   # As an example, we'll look at the famous Ishigami function
   # (A and B are specific parameters for the Ishigami function)
   A = 7
   B = 0.1
   y = np.arcsin(((n1*FF+n2-FF*n2)*x2-x1)/x2)
   y_{deg} = 180/np.pi * y
   # ====== TO HERE and replace with your own piece of code
   return y_deg
def conv_study(n, Y, X):
   # take n samples among the num samples, without replacement
   subset = np.random.choice(num_samples, size=n, replace=False)
   return rbd_fast.analyze(problem=problem,
                          Y=Y[subset],
                          X=X[subset])['S1']
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```
def bootstrap(problem, Y, X):
              Calculate confidence intervals of rbd-fast indices
              1000 draws
              returns 95% confidence intervals of the 1000 indices
              problem : dictionnary as SALib uses it
              X : SA input(s)
              Y : SA output(s)
              all_indices = []
              for i in range(1000):
                  X_{new} = np.zeros(X.shape)
                  Y_new = np.zeros(Y.shape).flatten()
                  # draw with replacement
                  tirage_indices = np.random.randint(0, high=Y.shape[0], size = Y.shape[0])
                  for j, index in enumerate(tirage_indices):
                      X_{new[j,:]} = X[index,:]
                      Y_{new[j]} = Y[index]
                  all_indices.append(rbd_fast.analyze(problem=problem, Y=Y_new, X=X_new)['S1']
              means = np.array([i.mean() for i in np.array(all_indices).T])
              stds = np.array([i.std() for i in np.array(all_indices).T])
              return np.array([means - 2 * stds, means + 2 * stds])
In [10]:
          # STATE THE PROBLEM DICTIONNARY
          # what will be varying (=inputs) ? in what bounds ?
          g_period_power = (np.math.pow(10,-9)) # structure grating period is 660 nanometers
          g_period_min = 620 * g_period_power # 620 nm
          g_period_max = 690 * g_period_power # 690 nm
          lmbd_BAND_C_min = 1530*g_period_power
          lmbd_BAND_C_max = 1565*g_period_power
          problem = {
              'num vars': 2,
              'names': ['Wavelength', 'GratingPeriod'],
              'bounds': [[lmbd_BAND_C_min,lmbd_BAND_C_max ],
                         [g_period_min, g_period_max]]
          # say we want to draw 150 samples
          num samples = 150
          # we draw a Latin Hypercube Sampling (LHS) that is fitted for an RBD FAST analysis
          # (other sampling metods available in the library though)
          from SALib.sample.latin import sample
          all_samples = sample(problem, num_samples)
          # run your model, procedure, experiment for each sample of your sampling
          # unpack all_samples into 3 vectors x1, x2, x3
          x1, x2 = all samples.T
```

```
In [11]: # run the model, all samples at the same time
    ishigami_results = run_model(x1, x2)
```

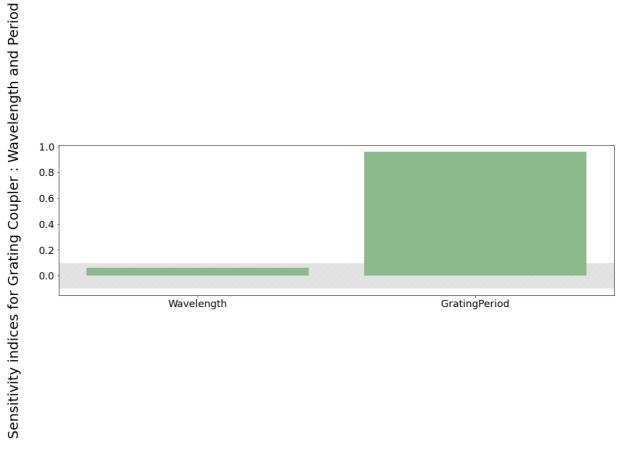
```
# Let us look at the analyze method

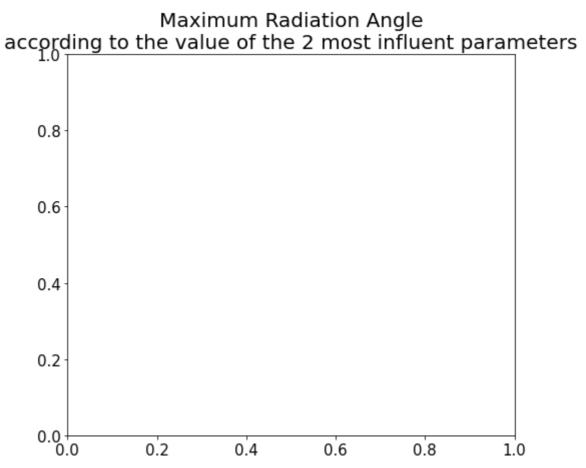
rbd_fast.analyze(problem=problem, Y=ishigami_results, X=all_samples)
# storing the first order indices of the analyze method
si1 = rbd_fast.analyze(problem=problem, Y=ishigami_results, X=all_samples)['S1']

# make nice plots with the indices (looks good on your presentations)
# do not use the plotting tools of SALib, they are made for the method of Morris ...
```

```
In [12]:
          import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
          from SALib.analyze import rbd_fast
          # this is where you define your own model, procedure, experiment...
          from SALib.test_functions import Ishigami
          import numpy as np
          fig, ax = plt.subplots()
          fig.set_size_inches(18, 5)
          ax.tick_params(labelsize=18)
          # ===== X-AXIS =====
          ax.set_xticks(np.arange(problem['num_vars']))
          ax.set_xticklabels(problem['names'])
          ax.set_xlim(xmin=-0.5, xmax=problem['num_vars'] - 0.5)
          # ===== Y-AXIS =====
          ax.set_ylabel('Sensitivity indices for Grating Coupler : Wavelength and Period\n', f
          # ==== BARS REPRESENTING THE SENSITIVITY INDICES =====
          ax.bar(np.arange(problem['num_vars']), si1,
                 color='DarkSeaGreen');
          #in striped grey : not significant indices
          ax.fill_between(x=[-0.5, 5.5], y1=-0.1, y2=0.1, color='grey', alpha=0.2, hatch='//',
          # take a closer look to undestand interactions (looks even better on your presentati
          # this part can be done without analyzing with SALib, just with the ouput from the s
          fig, ax = plt.subplots()
          fig.set_size_inches(8, 7)
          ax.tick params(labelsize=15)
          ax.set_title('Maximum Radiation Angle\n'
                        'according to the value of the 2 most influent parameters',
                       fontsize=20)
```

Out[12]: Text(0.5, 1.0, 'Maximum Radiation Angle\naccording to the value of the 2 most influen t parameters')





In [13]:	#######################################
In []:	

In []:	
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