SPE_HW2_Davide_Deborah

April 23, 2024

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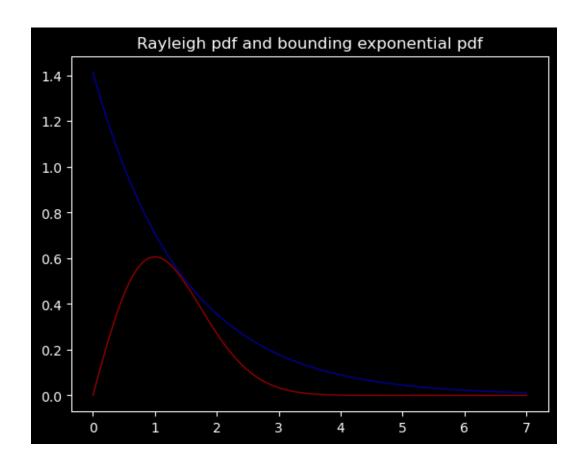
1 Exercise 1

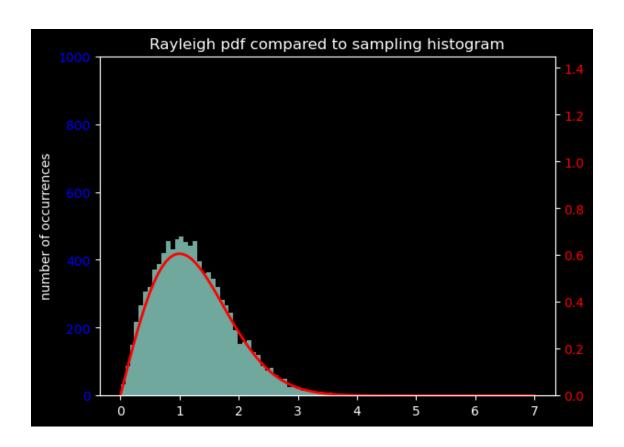
```
[]: import numpy as np
     from scipy import stats
     from scipy.stats import rayleigh
     from scipy.stats import expon
     from scipy.special import i0
     import random
     from random import uniform
     import matplotlib.pyplot as plt
     import math
     import numpy as np
     sz = 20000
     bin_default = 80
     #----- COMMON FUNCTIONS -----
     #calculated the value of the exponent function
     def calculate_expon(x, sc, lambd):
        return sc*lambd*math.e**(-lambd*x)
     \#rejection\ sampling,\ samples = x\ values,\ y1 = bounding\ funtion,\ y2 = function_{\square}
     ⇔to sample
     def accept(samples,y1,y2):
        accepted = []
        for a,b,c in zip(samples, y1, y2):
            ran = random.uniform(0,b)
             if ran <=c:</pre>
                 accepted.append(a)
        return accepted
```

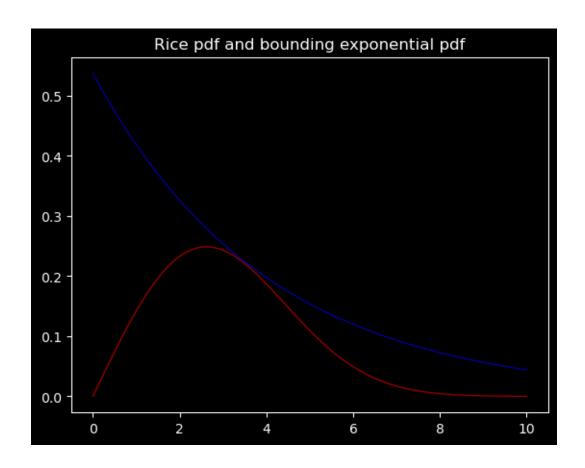
```
#plots the function and the histogram
def plot_comparison(accepted, x, y, name, ylim1, ylim2, bin_num=bin_default):
   fig, ax1 = plt.subplots()
   ax1.hist(accepted, bins=bin_num, alpha=0.8)
   plt.ylim(0,ylim1)
   plt.ylabel("number of occurrences")
   for tl in ax1.get_yticklabels():
       tl.set color('blue')
   ax2 = ax1.twinx()
   ax2.plot(x, y,'r-', lw=2, alpha=1, label='rice pdf')
   plt.ylim([0,ylim2])
   for tl in ax2.get_yticklabels():
       tl.set_color('red')
   plt.title(name+" pdf compared to sampling histogram")
# function to calculate Rayleigh pdf at point x
def calculate_ray(x,sigma):
   return (x/(sigma**2))*math.e**(-x**2/(2*sigma**2))
\# function to calculate Rice pdf at point x
def calculate rice(x):
   sigma=2
   v=2
   return (x/sigma**2)*math.e**(-(x**2 + v**2)/(2*sigma**2))*i0(x*v/sigma**2)
# ----- RAYLEIGH -----
# generating rayleigh and exponential pdf
sc=2.05
scale_factor=1.45
lambd=1/scale_factor
fig, ax = plt.subplots(1, 1)
x = np.linspace(0,7, 100)
ray = rayleigh()
y_ray = [calculate_ray(i,1) for i in x]
y_expon = [calculate_expon(i, sc, lambd) for i in x]
# plotting ray and exponential
exp = expon(scale=scale_factor)
ax.plot(x, y_expon,'b-', lw=1, alpha=0.6, label='expon pdf')
ax.plot(x, y_ray, 'r-', lw=1, alpha=0.6, label='rayleigh pdf')
plt.title("Rayleigh pdf and bounding exponential pdf")
```

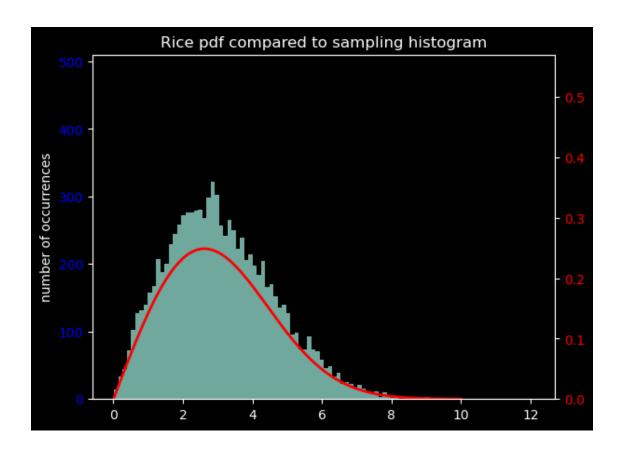
```
# generating samples according to exponential pdf
samples=exp.rvs(size=sz)
#calculating exp and ray values for the samples
exp_y_samples = [calculate_expon(i, sc, lambd) for i in samples]
ray_y_samples = [calculate_ray(i,1) for i in samples]
#accepting values
accepted = accept(samples, exp_y_samples, ray_y_samples)
#plotting comparison between rayleigh pdf and accepted values histogram
plot_comparison(accepted, x, y_ray, "Rayleigh", 1000, 1.45, bin_num = 60)
print("Samples per accept Ray:",sz/len(accepted))
# ------ RICE -----
# generating and plotting exponential bounding pdf and rice pdf
sc=2.15
scale_factor=4
lambd=1/scale_factor
fig, ax = plt.subplots(1, 1)
x = np.linspace(0,10,200)
exp_2 = expon(scale = scale_factor)
y_rice = [calculate_rice(i) for i in x]
y_expon = [calculate_expon(i,sc,lambd) for i in x]
ax.plot(x, y_rice,'r-', lw=1, alpha=0.6, label='rice pdf')
ax.plot(x, y_expon, 'b-', lw=1, alpha=0.6, label='exponential pdf')
plt.title("Rice pdf and bounding exponential pdf")
# sampling exponential bounding pdf
samples=exp_2.rvs(size=sz)
# caculating rice and exp. values for the samples
y expon samples = [calculate_expon(i,sc,lambd) for i in samples]
y_rice_samples = [calculate_rice(i) for i in samples]
#accepting values
accepted=[]
accepted = accept(samples, y_expon_samples, y_rice_samples)
#plotting comparison between rice pdf and accepted value histogram
plot_comparison(accepted, x, y_rice, "Rice", 510, 0.57, bin_num=100)
print("Samples per accept Rice:",sz/len(accepted))
```

Samples per accept Ray: 2.045617265009717 Samples per accept Rice: 2.15726458850178







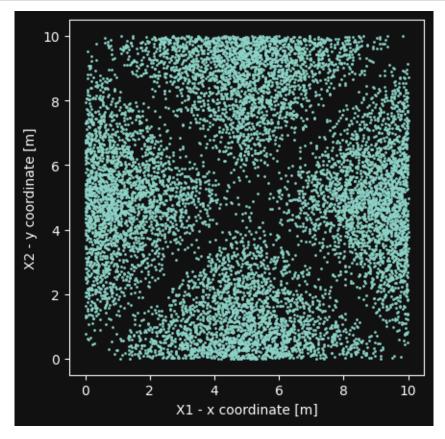


2 Exercise 2

2.1 1

```
samples = [sample_vector() for n in range(num_samples)]

# Plot the samples
plt.scatter([x[0] for x in samples], [x[1] for x in samples], s=1)
plt.xlabel('X1 - x coordinate [m]')
plt.ylabel('X2 - y coordinate [m]')
ax = plt.gca()
ax.set_aspect('equal')
plt.show()
```



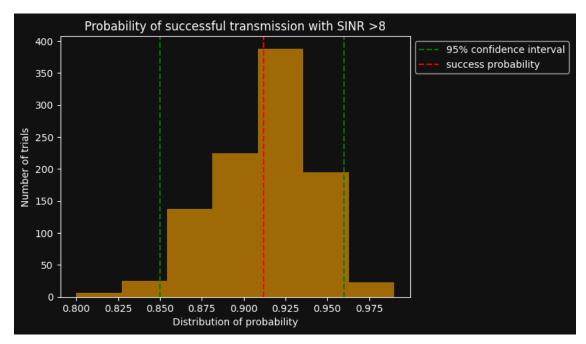
2.2 2

```
[]: import random
import matplotlib.pyplot as plt
import numpy as np
import seaborn as sns

# Constants
P_tx = 0.1
N = 1.6e-4
num_samples = 10000
```

```
num_trials = 1000 # number of trials
num_pairs = 100 # number of pairs per trial
random.seed(42)
def compute_distance(point1, point2):
    return np.sqrt((point1[0] - point2[0])**2 + (point1[1] - point2[1])**2)
def compute_snr(d_tx_rx):
    return (P_tx / (N * d_tx_rx**2))
def simulate trial():
    success_count = 0
    for n in range(num_pairs):
        tx_index = random.randint(0, num_samples-1)
        rx_index = random.randint(0, num_samples-1)
        while rx_index == tx_index:
            rx_index = random.randint(0, num_samples-1)
        d_tx_rx = compute_distance(samples[tx_index], samples[rx_index])
        snr = compute_snr(d_tx_rx)
        if snr > 8:
            success_count += 1
    return success_count / num_pairs
def sample vector():
    while True:
        x1 = random.uniform(0, 10)
        x2 = random.uniform(0, 10)
        u = random.uniform(0, 10)
        if u \le abs((-1)*x1 +10 - x2):
            if u \le abs(x1-x2):
                break
    return (x1, x2)
samples = [sample_vector() for n in range(num_samples)]
success_probabilities = []
trials = []
for n in range(num_trials):
    trial = simulate trial()
    success_probabilities.append(trial)
    trials.append(n)
success_probability = np.mean(success_probabilities)
confidence_interval = np.percentile(success_probabilities, [2.5, 97.5])
```

Success probability: 0.9120 95% confidence interval: [0.8500, 0.9600]



2.3 3

```
[]: import random
  import matplotlib.pyplot as plt
  import numpy as np

# Constants
```

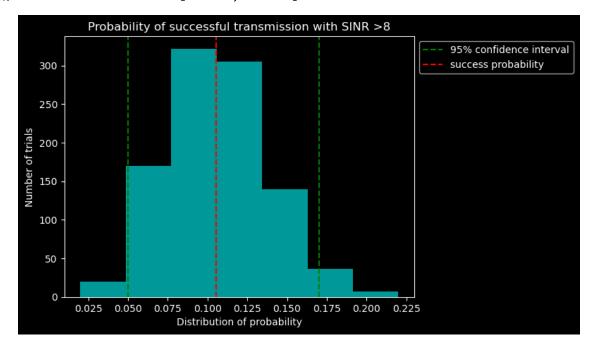
```
P_tx = 0.1
N = 1.6e-4
num_samples = 10000
num_trials = 1000 # number of trials
num_pairs = 100 # number of pairs per trial
random.seed(42)
def compute_distance(point1, point2):
         return np.sqrt((point1[0] - point2[0])**2 + (point1[1] - point2[1])**2)
def compute_sinr(d_tx_rx, d_txI_rx):
         return ((P_tx*d_tx_rx**(-2)) / (N + P_tx*d_txI_rx**(-2)))
def simulate_trial():
         success_count = 0
         for n in range(num_pairs):
                   tx_index = random.randint(0, num_samples-1)
                   rx_index = random.randint(0, num_samples-1)
                   txI_index = random.randint(0, num_samples-1)
                   while rx_index == tx_index or txI_index == tx_index or rx_index ==_u
   →txI_index:
                             rx_index = random.randint(0, num_samples-1)
                             txI_index = random.randint(0, num_samples-1)
                   d_tx_rx = compute_distance(samples[tx_index], samples[rx_index])
                   d_txI_rx = compute_distance(samples[tx_index], samples[txI_index])
                   sinr = compute_sinr(d_tx_rx, d_txI_rx)
                   if sinr > 8:
                              success_count += 1
                              \#print(" d:"+str(d_tx_rx)+", di:"+str(d_txI_rx)+" - d/di:"+str(d_txI_rx)+" - d/di:"+str(d_txI_
  \rightarrow"+str(d tx rx/d txI rx) + "--> d<di:"+str(bool(d tx rx<d txI rx)))#+" -- se_1

→: "+ str(bool(d_txI_rx-d_tx_rx>2)))
         return success_count / num_pairs
def sample vector():
         while True:
                   x1 = random.uniform(0, 10)
                   x2 = random.uniform(0, 10)
                   u = random.uniform(0, 10)
                   if u \le abs((-1)*x1 +10 - x2):
                              if u \le abs(x1-x2):
                                       break
         return (x1, x2)
samples = [sample_vector() for n in range(num_samples)]
```

```
success_probabilities = []
for n in range(num_trials):
   trial = simulate_trial()
    success_probabilities.append(trial)
success_probability = np.mean(success_probabilities)
confidence_interval = np.percentile(success_probabilities, [2.5, 97.5])
print(f"Success probability: {success_probability:.4f}")
print(f"95% confidence interval: [{confidence_interval[0]:.4f},__

√{confidence_interval[1]:.4f}]")
plt.hist(success_probabilities, color='cyan', alpha=0.6, bins=7)
plt.xlabel('Distribution of probability')
plt.ylabel('Number of trials')
plt.axvline(x=confidence_interval[0], color="green", ls = "--", label="95%__
 ⇔confidence interval")
plt.axvline(x=confidence_interval[1], color="green", ls="--")
plt.axvline(x=success_probability, color="red", ls="--", label="success_"
 ⇔probability")
plt.title( "Probability of successful transmission with SINR >8")
plt.legend(bbox_to_anchor=(1.0, 1), loc='upper left')
plt.show()
```

Success probability: 0.1055 95% confidence interval: [0.0500, 0.1700]



2.4 4

```
[]: import random
     import numpy as np
     # Constants
     P tx = 0.1
     N = 1.6e-4
     num_samples = 10000
    num_sample_exp = 1000
     num_trials = 1000 # number of trials
     num_pairs = 100 # number of pairs per trial
     random.seed(42)
     # exponential variable tx
     e_mean = 1 \# 1/lambda
     def sample_exponential(mean):
         return -mean * np.log(1 - random.random())
     def exp_def():
         exp_samples = [sample_exponential(e_mean) for n in range(num_sample_exp)]
         e_index = random.randint(0, num_sample_exp-1)
         e = exp_samples[e_index]
         return e
     def compute_distance(point1, point2):
         return np.sqrt((point1[0] - point2[0])**2 + (point1[1] - point2[1])**2)
     def compute_snr_f(d_tx_rx,e_tx):
         return (P_tx * e_tx/ (N * d_tx_rx**2))
     def compute_sinr_f(d_tx_rx, d_txI_rx, e_tx, e_I):
         return ((P_tx*e_tx*d_tx_rx**(-2)) / (N + P_tx*e_I*d_txI_rx**(-2)))
     def simulate_trial_snr_f():
         success_count = 0
         for n in range(num_pairs):
             tx_index = random.randint(0, num_samples-1)
             rx_index = random.randint(0, num_samples-1)
             while rx_index == tx_index:
                 rx_index = random.randint(0, num_samples-1)
             d_tx_rx = compute_distance(samples[tx_index], samples[rx_index])
             snr = compute_snr_f(d_tx_rx, exp_def())
```

```
if snr > 8:
            success_count += 1
   return success_count / num_pairs
def simulate_trial_sinr_f():
   success_count = 0
   for n in range(num_pairs):
       tx_index = random.randint(0, num_samples-1)
       rx index = random.randint(0, num samples-1)
       txI_index = random.randint(0, num_samples-1)
       while rx_index == tx_index or txI_index == tx_index or rx_index ==_u
 ⇔txI_index:
           rx_index = random.randint(0, num_samples-1)
           txI_index = random.randint(0, num_samples-1)
       d_tx_rx = compute_distance(samples[tx_index], samples[rx_index])
       d_txI_rx = compute_distance(samples[tx_index], samples[txI_index])
       sinr = compute_sinr_f(d_tx_rx, d_txI_rx, exp_def(),exp_def())
        if sinr > 8:
           success count += 1
   return success_count / num_pairs
def sample vector():
   while True:
       x1 = random.uniform(0, 10)
       x2 = random.uniform(0, 10)
       u = random.uniform(0, 10)
       if u \le abs((-1)*x1 +10 - x2):
           if u \le abs(x1-x2):
               break
   return (x1, x2)
samples = [sample_vector() for n in range(num_samples)]
ax1, ax2 = plt.subplots()
success_probabilities_snr = []
for n in range(num trials):
   trial = simulate_trial_snr_f()
    success_probabilities_snr.append(trial)
success_probability_snr = np.mean(success_probabilities_snr)
confidence_interval_snr = np.percentile(success_probabilities_snr, [2.5, 97.5])
print(f"SNR:Success probability: {success_probability_snr:.4f}")
print(f"SNR:95% confidence interval: [{confidence interval snr[0]:.4f},__
```

```
success_probabilities_sinr = []
for n in range(num_trials):
    trial = simulate_trial_sinr_f()
    success_probabilities_sinr.append(trial)
success_probability_sinr = np.mean(success_probabilities_sinr)
confidence_interval_sinr = np.percentile(success_probabilities_sinr, [2.5, 97.
 print(f"SINR:Success probability: {success probability sinr:.4f}")
print(f"SINR:95% confidence interval: [{confidence interval sinr[0]:.4f}, __

√{confidence_interval_sinr[1]:.4f}]")
plt.figure(0)
plt.hist(success_probabilities_snr, color='orange', alpha=0.6, bins=7)
plt.xlabel('Distribution of probability')
plt.ylabel('Number of trials')
plt.axvline(x=confidence_interval_snr[0], color="green", ls = "--", label="95%__
 ⇔confidence interval")
plt.axvline(x=confidence_interval_snr[1], color="green", ls="--")
plt.axvline(x=success_probability_snr, color="red", ls="--", label="success_"
 →probability")
plt.title( "Probability of successful transmission with SNR >8 and a fading ∪
 ⇔effect")
plt.legend(bbox_to_anchor=(1.0, 1), loc='upper left')
plt.show
plt.figure(1)
plt.hist(success_probabilities_sinr, color='cyan', alpha=0.6, bins=7)
plt.xlabel('Distribution of probability')
plt.ylabel('Number of trials')
plt.axvline(x=confidence_interval_sinr[0], color="green", ls = "--", label="95%"
  ⇔confidence interval")
plt.axvline(x=confidence_interval_sinr[1], color="green", ls="--")
plt.axvline(x=success_probability_sinr, color="red", ls="--", label="success_u
 →probability")
plt.title( "Probability of successful transmission with SINR >8 and a fading ∪
 ⇔effect")
plt.legend(bbox_to_anchor=(1.0, 1), loc='upper left')
plt.show
SNR:Success probability: 0.6764
SNR:95% confidence interval: [0.5800, 0.7602]
SINR: Success probability: 0.1617
SINR:95% confidence interval: [0.1000, 0.2400]
```

[]: <function matplotlib.pyplot.show(close=None, block=None)>

