```
In [124]:
```

```
import pandas as pd
df1 = pd.read_fwf('a.txt', encoding = "utf_16_le", engine = 'python', header = None)
```

In [125]:

```
df1.columns =['a']
#df1.astype('int32')
```

In [171]:

```
df2 = pd.read_fwf('b.txt', encoding = "utf_16_le", engine = 'python', header = None)
```

In [172]:

df2

Out[172]:

	0
0	0
1	20
2	92
3	48
4	16
9995	59
9996	83
9997	97
9998	85
9999	21

10000 rows × 1 columns

In [160]:

```
df2.columns =['b']
```

In [129]:

```
import numpy as np
from sklearn.preprocessing import MinMaxScaler
import matplotlib.pyplot as plt
%matplotlib inline
from scipy import stats
import math
import sys
```

In [130]:

```
df1.dropna(axis=0)
#df1.replace(0,np.nan, inplace = True)
```

Out[130]:

10000 rows × 1 columns

In [173]:

df2

Out[173]:

	0
0	0
1	20
2	92
3	48
4	16
9995	59
9996	83
9997	97
9998	85
9999	21
10000	rowo v

10000 rows × 1 columns

```
In [175]:
```

```
df2.dropna(axis =0)
df2.replace(0,np.nan, inplace=True)
#df2.astype('int64')
```

In [185]:

```
df2 =df2.where(pd.notnull(df2), df2.mean(), axis='columns').astype('int64')
```

In [186]:

```
# case1.df2의 시계열 결측값 0은 연속함수로 0이나 NaN으로 대체할 수없으므로 데이터간의 평균 값으로 대체하
#case2. 누적함수를 그리다보니 이전값 대체는 선형함수를 만들어내는 것 같아서 a 행의 값으로 결측값을 대체해
# df2.fillna(method= 'bfill')
# df2.replace(0,np.nan, inplace = True)
# df2['b'] = np.where(pd.isnull(df2['b'])==True, df1['a'], df2['b'])
```

In [191]:

```
df2.columns =['b']
df2
```

Out[191]:

```
b
0 49
1 20
2 92
3 48
4 16
...
9995 59
9996 83
9997 97
9998 85
9999 21
```

10000 rows × 1 columns

In [53]:

```
# new_df1= df1.dropna(axis=0)
# new_df1 = new_df1.fillna(method = 'ffill')
# new_df2 = df2.dropna(axis=0)
# new_df2 = new_df2.fillna(method = 'ffill')
```

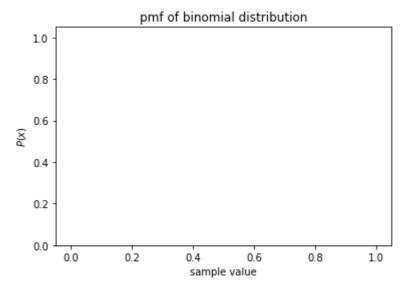
In [192]:

```
np_df1 = df1.to_numpy()
```

```
In [193]:
np_df2 =df2.to_numpy().astype('int')
In [194]:
x_range = np.arange(1,100,1)
In [195]:
y_range = np_df1
In [196]:
df2.rolling(window=10000)
df2.mean()
Out[196]:
b 49.4959
dtype: float64
In [197]:
import pylab as plt
N = x_range
X = y_range
counts, bins = np.histogram(X,bins=50, density=True)
bins = bins[:-1] + (bins[1] - bins[0])/2
print (np.trapz(counts, bins))
0.9997000000000001
In [198]:
df1.mean()
Out[198]:
a 49.9491
dtype: float64
In [199]:
df2.mean()
Out[199]:
b 49.4959
dtype: float64
In [200]:
N= 10000
mu = 49.9491
rv = stats.binom(N,mu)
```

In [201]:

```
xx = np .arange ( N + 1 )
plt.bar ( xx, rv.pmf(xx), align = "center" )
plt.xlabel ( "sample value" )
plt.ylabel ( "$ P (x) $" )
plt.title ( "pmf of binomial distribution" )
plt.show ()
```



In [65]:

```
df2 = df2.astype('int')
```

In [202]:

df1

Out[202]:

	а
0	52
1	52
2	56
3	59
4	52
9995	55
9996	47
9997	43
9998	57
9999	48

10000 rows × 1 columns

In [203]:

df2

Out[203]:

	b
0	49
1	20
2	92
3	48
4	16
9995	59
9996	83
9997	97
9998	85
9999	21

10000 rows × 1 columns

In [204]:

```
ab_df = pd.concat([df1,df2], axis = 1)
```

In [205]:

ab_df

Out[205]:

	а	b
0	52	49
1	52	20
2	56	92
3	59	48
4	52	16
9995	55	59
9996	47	83
9997	43	97
9998	57	85
9999	48	21

10000 rows × 2 columns

In [206]:
ab_df.mean()
Out[206]:
a 49.9491 b 49.4959 dtype: float64
In [207]:
ab_df['a'].median()
Out[207]:
50.0
In [208]:
ab_df['a'].mean()
Out[208]:
49.9491
In [209]:
ab_df['b'].mean()
Out[209]:
49.4959
In [210]:
ab_df['b'].median()
Out[210]:
49.0

In [211]:

ab_df

Out[211]:

	а	b	
0	52	49	
1	52	20	
2	56	92	
3	59	48	
4	52	16	
9995	55	59	
9996	47	83	
9997	43	97	
9998	57	85	
9999	48	21	

10000 rows × 2 columns

In [212]:

 $np_ab_df = np.concatenate([np_df1, np_df2], axis = 0)$

In [213]:

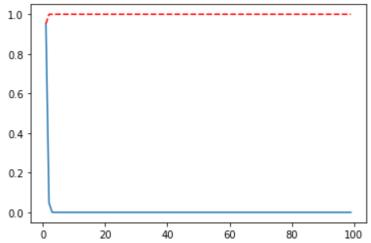
```
import numpy as np
from pylab import *

# Create some test data
dx = 1
X = np.arange(1, 100, dx)
Y = exp(-X ** 2)

# Normalize the data to a proper PDF
Y /= (dx * Y).sum()

# Compute the CDF
CY = np.cumsum(Y * dx)

# Plot both
plot(X, Y)
plot(X, CY, 'r--')
show()
```



In [346]:

Out[346]:

```
array([[52],
[52],
[56],
...,
[97],
[85],
[21]])
```

In [214]:

```
# define distributions
events = ['a','b']
p = np_df1
q = np_df2
print('P=%.3f Q=%.3f' % (sum(p), sum(q)))
```

In [215]:

```
# from matplotlib import pyplot

# # plot first distribution

# pyplot.subplot(2,1,1)

# pyplot.bar(events, p)

# # plot second distribution

# pyplot.subplot(2,1,2)

# pyplot.bar(events, q)

# # show the plot

# pyplot.show()
```

In [216]:

```
# example of calculating the kl divergence between two mass functions
from math import log2
def kl_divergence(p,q):
    return sum(p[i] * log2(p[i]/q[i]) for i in range(len(p)))
```

In [217]:

```
# calculate (P || Q)
kl_pq = kl_divergence(p, q)
print('KL(P || Q): %.3f bits' % kl_pq)
# calculate (Q || P)
kl_qp = kl_divergence(q, p)
print('KL(Q || P): %.3f bits' % kl_qp)
```

KL(P || Q): 222497.557 bits KL(Q || P): 139015.771 bits

/home/yjoh/.conda/envs/test/lib/python3.7/site-packages/ipykernel_launcher.py:4: Deprecation Warning: Calling np.sum(generator) is deprecated, and in the future will give a different result. Use np.sum(np.fromiter(generator)) or the python sum builtin instead. after removing the cwd from sys.path.

In [218]:

```
# 8. a) different
# KL(P || Q) != KL(Q || P)
# KL divergence score is not symmetrical.
# it also referred to as "relative entropy" (Pattern recognition and machine learning, 2006)
# 출처:https://machinelearningmastery.com/divergence-between-probability-distributions/
```

In [219]:

```
# example of calculating the kl divergence (relative entropy) with scipy

from scipy.special import rel_entr

# define distributions

p = np_df1

q = np_df2

# calculate (P || Q)

kl_pq = rel_entr(p, q)

print('KL(P || Q): %.3f nats' % sum(kl_pq))

# calculate (Q || P)

kl_qp = rel_entr(q, p)

print('KL(Q || P): %.3f nats' % sum(kl_qp))
```

KL(P || Q): 154223.554 nats KL(Q || P): 96358.389 nats

In [220]:

```
# calculate the js divergence
def js_divergence(p, q):
   m = 0.5 * (p + q)
   return 0.5 * kl_divergence(p, m) + 0.5 * kl_divergence(q, m)
```

In [221]:

```
from math import log2
from math import sqrt
from numpy import asarray
# calculate JS(P || Q)
js_pq = js_divergence(p, q)
print('JS(P || Q) divergence: %.3f bits' % js_pq)
print('JS(P || Q) distance: %.3f' % sqrt(js_pq))
```

JS(P || Q) divergence: 40038.861 bits JS(P || Q) distance: 200.097

/home/yjoh/.conda/envs/test/lib/python3.7/site-packages/ipykernel_launcher.py:4: Deprecation Warning: Calling np.sum(generator) is deprecated, and in the future will give a different result. Use np.sum(np.fromiter(generator)) or the python sum builtin instead. after removing the cwd from sys.path.

In [222]:

```
# calculate JS(Q || P)
js_qp = js_divergence(q, p)
print('JS(Q || P) divergence: %.3f bits' % js_qp)
print('JS(Q || P) distance: %.3f' % sqrt(js_qp))
```

JS(Q || P) divergence: 40038.861 bits

JS(Q || P) distance: 200.097

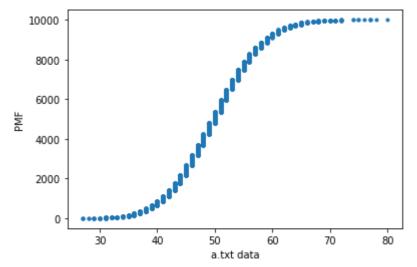
/home/yjoh/.conda/envs/test/lib/python3.7/site-packages/ipykernel_launcher.py:4: Deprecation Warning: Calling np.sum(generator) is deprecated, and in the future will give a different result. Use np.sum(np.fromiter(generator)) or the python sum builtin instead. after removing the cwd from sys.path.

In [223]:

We can see that the calculation is symmetrical, giving the same score and distance measure for JS(P || Q) and

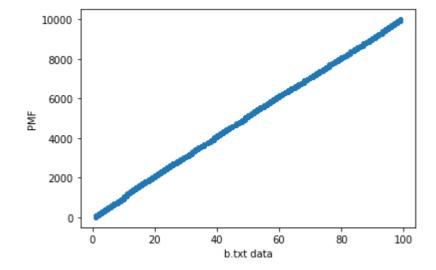
In [224]:

```
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
x_range = np.sort(ab_df['a'])
y_range = np.arange(1,len(x_range)+1/len(x_range))
_ = plt.plot(x_range,y_range, marker = '.', linestyle = 'none')
_ = plt.xlabel('a.txt data')
_ = plt.ylabel('PMF')
plt.margins(0.05)
plt.show()
```



In [225]:

```
x_range = np.sort(ab_df['b'])
y_range = np.arange(1,len(x_range)+1/len(x_range))
_ = plt.plot(x_range,y_range, marker = '.', linestyle = 'none')
_ = plt.xlabel('b.txt data')
_ = plt.ylabel('PMF')
plt.margins(0.05)
plt.show()
```



In [226]:

```
#plot cdf of a.txt
counts, bin_edges = np.histogram(ab_df['a'], bins = 10, density = True)
pdf = counts / (sum(counts))
print(pdf);
print(bin_edges)
```

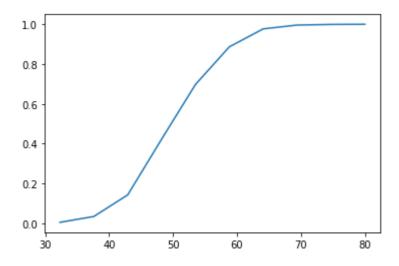
[0.005 0.0294 0.1084 0.2798 0.275 0.1891 0.0903 0.0191 0.0031 0.0008] [27. 32.3 37.6 42.9 48.2 53.5 58.8 64.1 69.4 74.7 80.]

In [227]:

```
#compute cdf
cdf = np.cumsum(pdf)
#plt.plot(bin_edges[1:],pdf)
plt.plot(bin_edges[1:],cdf)
```

Out[227]:

[<matplotlib.lines.Line2D at 0x7f22ce058690>]



In [228]:

```
#plot cdf of b.txt
counts, bin_edges = np.histogram(ab_df['b'], bins = 10, density = True)
pdf = counts / (sum(counts))
print(pdf);
print(bin_edges)
```

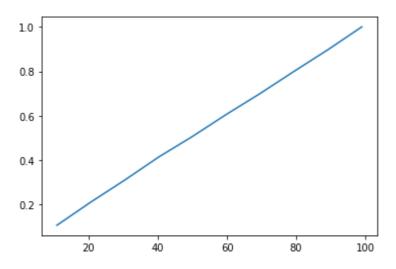
[0.1069 0.1043 0.0989 0.1036 0.0929 0.0992 0.0944 0.0996 0.0974 0.1028] [1. 10.8 20.6 30.4 40.2 50. 59.8 69.6 79.4 89.2 99.]

In [229]:

```
#compute cdf
cdf = np.cumsum(pdf)
#plt.plot(bin_edges[1:],pdf)
plt.plot(bin_edges[1:],cdf)
```

Out[229]:

[<matplotlib.lines.Line2D at 0x7f22ce240310>]



In [230]:

```
def entropy(x, bins=None):
    x = ab_df['a']
    N = x.shape[0]
    if bins is None:
        counts = np.bincount(x)
    else:
        counts = np.histogram(x, bins=bins)[0] # 0th idx is counts
    p = counts[np.nonzero(counts)]/N # avoids log(0)
    H = -np.dot( p, np.log2(p) )
    return H
```

In [231]:

```
entropy(ab_df['a'])
```

Out[231]:

4.848677778718739

```
In [232]:
```

```
def entropy(x, bins=None):
    x = ab_df['b']
    N = x.shape[0]
    if bins is None:
        counts = np.bincount(x)
    else:
        counts = np.histogram(x, bins=bins)[0] # 0th idx is counts
    p = counts[np.nonzero(counts)]/N # avoids log(0)
    H = -np.dot( p, np.log2(p) )
    return H
```

In [233]:

```
entropy(ab_df['b'])
```

Out[233]:

6.61813855112409

In [234]:

```
def entropy(x, bins=None):
    x = ab_df
    N = x.shape[0]
    if bins is None:
        counts = np.bincount(x)
    else:
        counts = np.histogram(x, bins=bins)[0] # 0th idx is counts
    p = counts[np.nonzero(counts)]/N # avoids log(0)
    H = -np.dot( p, np.log2(p) )
    return H
```

In [235]:

```
periods = 20
tidx = pd.date_range(0, periods = periods, freq = 'T')
data = np.random.randn(periods)
ts = pd.Series(data = data, index = tidx, name = 'random sampling=20')
ts.describe()
```

Out[235]:

```
20.000000
count
mean
        0.428488
      0.839414
std
      -1.645001
min
25%
       0.089325
50%
       0.585502
75%
       0.794602
       1.769231
max
Name: random sampling=20, dtype: float64
```

In [252]:

```
#시계열 샘플 n= 20 추출
time_series_n_get_sample = ab_df.loc[[20,21,22,23,24,25,26,27,28,29,30,31,32,33,34,35,36,37,38,39],:]
```

In [254]:

time_series_n_get_sample.head(20)

Out[254]:

	а	b
20	46	73
21	45	81
22	46	41
23	48	8
24	58	24
25	53	7
26	53	78
27	51	9
28	58	13
29	46	72
30	45	62
31	50	89
32	55	58
33	44	28
34	56	76
35	54	41
36	30	10
37	53	89
38	62	22
39	56	68

In [260]:

```
#n = 20 2 time series n get sample 2 cross entropy H(P,Q) 구하기
def cross_entropy(x, y):
  """ Computes cross entropy between two distributions.
  Input: x: iterabale of N non-negative values
       y: iterabale of N non-negative values
  Returns: scalar
  if np.any(x < 0) or np.any(y < 0):
    raise ValueError('Negative values exist.')
  # Force to proper probability mass function.
  x = np.array(x, dtype=np.float)
  y = np.array(y, dtype=np.float)
  x = np.sum(x)
  y = np.sum(y)
  # Ignore zero 'y' elements.
  mask = y > 0
  x = x[mask]
  y = y[mask]
  ce = -np.sum(x * np.log(y))
  return ce
def cross entropy via scipy(x, y):
    "" SEE: https://en.wikipedia.org/wiki/Cross_entropy"
    return entropy(x) + entropy(x, y)
from scipy.stats import entropy, truncnorm
x = truncnorm.rvs(0.1, 2, size=100)
y = truncnorm.rvs(0.1, 2, size=100)
print (np.isclose(cross_entropy(x, y), cross_entropy_via_scipy(x, y)))
True
In [261]:
cross entropy(time series n get sample['a'],time series n get sample['b'])
Out[261]:
3.287789895407153
In [267]:
# define distributions
events = ['a','b']
P = time_series_n_get_sample['a'].to_numpy()
Q = time_series_n_get_sample['b'].to_numpy()
print('P=%.3f Q=%.3f' % (sum(P), sum(Q)))
```

P=1009.000 Q=949.000

In [268]:

```
# example of calculating the kl divergence between two mass functions

from math import log2

def kl_divergence(p,q):
    return sum(p[i] * log2(p[i]/q[i]) for i in range(len(p)))
```

In [269]:

```
# calculate (P || Q)

kl_pq = kl_divergence(P, Q)

print('KL(P || Q): %.3f bits' % kl_pq)

# calculate (Q || P)

kl_qp = kl_divergence(Q, P)

print('KL(Q || P): %.3f bits' % kl_qp)
```

KL(P || Q): 528.784 bits KL(Q || P): 221.243 bits

/home/yjoh/.conda/envs/test/lib/python3.7/site-packages/ipykernel_launcher.py:4: Deprecation Warning: Calling np.sum(generator) is deprecated, and in the future will give a different result. Use np.sum(np.fromiter(generator)) or the python sum builtin instead. after removing the cwd from sys.path.

In [270]:

```
#8. 전혀다르다.
# KL(P || Q) != KL(Q || P)
# KL divergence score is not symmetrical.
```

In [271]:

```
#9.  
# As such, we can calculate the cross-entropy by adding the entropy of the distribution plus the additional entropy  
# This is intuitive, given the definition of both calculations; for example:  
# H(P, Q) = H(P) + KL(P || Q)  
cross_entropy(p,q)
```

Out[271]:

9.500113212652087

In [272]:

```
#10. same.
# H(P, Q) = H(P) + KL(P || Q)
```

In [273]:

```
# calculate the js divergence

def js_divergence_sample(P, Q):

m = 0.5 * (P + Q)

return 0.5 * kl_divergence(P, m) + 0.5 * kl_divergence(Q, m)
```

In [274]:

```
# calculate JS(P || Q)
js_pq_sample = js_divergence(P, Q)
print('JS(P || Q) divergence: %.3f bits' % js_pq_sample)
print('JS(P || Q) distance: %.3f' % sqrt(js_pq_sample))
```

JS(P || Q) divergence: 86.535 bits JS(P || Q) distance: 9.302

/home/yjoh/.conda/envs/test/lib/python3.7/site-packages/ipykernel_launcher.py:4: Deprecation Warning: Calling np.sum(generator) is deprecated, and in the future will give a different result. Use np.sum(np.fromiter(generator)) or the python sum builtin instead. after removing the cwd from sys.path.

In [275]:

```
# calculate JS(Q || P)
js_qp_sample = js_divergence(Q, P)
print('JS(Q || P) divergence: %.3f bits' % js_qp_sample)
print('JS(Q || P) distance: %.3f' % sqrt(js_qp_sample))
```

JS(Q || P) divergence: 86.535 bits JS(Q || P) distance: 9.302

/home/yjoh/.conda/envs/test/lib/python3.7/site-packages/ipykernel_launcher.py:4: Deprecation Warning: Calling np.sum(generator) is deprecated, and in the future will give a different result. Use np.sum(np.fromiter(generator)) or the python sum builtin instead. after removing the cwd from sys.path.

In [276]:

13.same

In [277]:

#14. why cross entropy is widely used as a loss function in ML problems.

In []:

#pros of KL divergence as a distance matric: #cons of KL divergence as a distance matric: #why it is useful /not useful?