

## 2 × 2 分割表（クロス集計）とカイ2乗検定

```
In [1]: import matplotlib.pyplot as plt  
import numpy as np  
import pandas as pd  
from scipy import stats
```

### データの読み込み

```
In [2]: pd.read_csv("D:/2022_数理統計学/StatData/StatData11_3.csv",  
encoding='cp932').head() # csv ファイルの中をのぞき見
```

Out[2]:

試験の合否(PF) 過半数の出席(PA)		
0	P	A
1	F	P
2	F	P
3	F	A
4	F	P

```
In [3]: # 日本語を避ける  
Data=pd.read_csv("D:/2022_数理統計学/StatData/StatData11_3.csv",  
encoding='cp932',  
skiprows=1,  
names=('Exam', 'Attend'))  
Data.head()
```

Out[3]:

	Exam	Attend
0	P	A
1	F	P
2	F	P
3	F	A
4	F	P

### クロス集計

```
In [4]: Data_cross=pd.crosstab(Data['Exam'], Data['Attend']) # クロス集計  
Data_cross
```

Out[4]:

Attend	A	P
Exam		
F	23	46
P	15	41

```
In [5]: # 集計  
Data_cross.sum()
```

```
Out[5]: Attend  
A    38  
P    87  
dtype: int64
```

```
In [6]: Data_cross.sum(axis=1)
```

```
Out[6]: Exam  
F    69  
P    56  
dtype: int64
```

```
In [7]: # 集計（変数名を付けておく）  
Attend_A, Attend_P=Data_cross.sum()  
Exam_F, Exam_P=Data_cross.sum(axis=1)  
N=Attend_A+Attend_P # データ総数  
Attend_A, Attend_P, Exam_F, Exam_P, N
```

```
Out[7]: (38, 87, 69, 56, 125)
```

## 独立性を仮定したときの理論度数

```
In [8]: # Exam と Attend の独立性を仮定したときの理論度数  
Exam_F_Attend_A=N*(Exam_F/N)*(Attend_A/N)  
Exam_F_Attend_A
```

```
Out[8]: 20.976
```

```
In [9]: Exam_F_Attend_A=N*(Exam_F/N)*(Attend_A/N)  
Exam_F_Attend_P=N*(Exam_F/N)*(Attend_P/N)  
Exam_P_Attend_A=N*(Exam_P/N)*(Attend_A/N)  
Exam_P_Attend_P=N*(Exam_P/N)*(Attend_P/N)  
Exam_F_Attend_A, Exam_F_Attend_P, Exam_P_Attend_A, Exam_P_Attend_P
```

```
Out[9]: (20.976, 48.02399999999994, 17.024, 38.976)
```

## カイ2乗検定

```
In [10]: Expected=pd.DataFrame(  
    {'A': [Exam_F_Attend_A, Exam_P_Attend_A],  
     'P': [Exam_F_Attend_P, Exam_P_Attend_P]},  
    index=['F', 'P'])  
Expected
```

```
Out[10]:
```

	A	P
F	20.976	48.024
P	17.024	38.976

```
In [11]: # chi-square  
chi_square=((Data_cross-Expected)**2/Expected).sum().sum()  
chi_square
```

```
Out[11]: 0.6263413562642237
```

```
In [12]: # 棄却域  
alpha=0.05  
Chi=stats.chi2(1) # 自由度 1 のカイに乗分布  
Chi_alpha=Chi.isf(alpha) # 上側 alpha 点  
Chi_alpha
```

```
Out[12]: 3.8414588206941285
```

```
In [13]: # 棄却されるか?  
chi_square>Chi_alpha
```

```
Out[13]: False
```

```
In [14]: # P 値  
P=1-Chi.cdf(chi_square)  
P
```

```
Out[14]: 0.4287005129151674
```

```
In [ ]:
```