

```
library(reticulate)
use_python("/Users/che/pythonProject2/venv/bin/python")
```

```
knitr::opts_chunk$set(
  echo = TRUE,
  fig.align = "center",
  fig.path = "fig/",
  fig.showtext = TRUE,
  message = FALSE,
  warning = FALSE
)
```

## 作业说明

仿照二项分布的例子,试给出超几何分布、多项分布和多元超几何分布的例子,列出参考出处(书本文献,或者网页),并给出相应例子的R和Python实现.

- 提交作业截止时间: 2022-03-18-21:00 (周五)
- 打包发送QQ邮箱; 文件名和邮件名格式如: amsa-hw2-20191003699-renyifei.zip, 包含如:
  - amsa-hw2-20191003699-renyifei.md
  - amsa-hw2-20191003699-renyifei.rmd
  - amsa-hw2-20191003699-renyifei.pdf
  - amsa-hw2-20191003699-renyifei.html
- 注意事项
  - 打包文件zip里面一定要有pdf文件, 便于批注和批改
    - 可使用Typora编辑Markdown文件, 然后导出pdf, 再打包.md和.pdf文件
    - 亦可使用RStudio编辑R Markdown文件, 然后生成pdf, 再打包.rmd和.pdf文件
    - R Markdown转pdf有困难, 可导出html并设定 `keep_md: yes`, 再将md转pdf
      - 需要提供.rmd, .html, .md和.pdf, 将其打包上传

## 二项分布

如果随机变量有如下的概率分布:

$$P(X = k) = C_n^k p^k q^{n-k}, \quad k = 0, 1, \dots, n$$

则称  $X$  服从二项分布, 其中  $p, q > 0, p + q = 1$ , 记做  $X \sim B(n, p)$ .

称为二项分布的原因是  $C_n^k p^k q^{n-k}$  为二项展开式

$$(p + q)^n = \sum_{k=0}^n C_n^k p^k q^{n-k}$$

的第  $k + 1$  项. B 是 Binomial 的缩写.

## 例

说野外正在进行 9 ( $n = 9$ ) 口石油勘探井的发掘工作, 每一口井能够开发出油的概率是 0.1 ( $p = 0.1$ ). 请问, 最终所有的勘探井都勘探失败的概率?

我们手动用公式计算得:

$$\binom{n}{k} p^k (1 - p)^{n-k} = \binom{9}{0} 0.1^0 (1 - 0.1)^9 = 0.1^9 \approx 0.3874$$

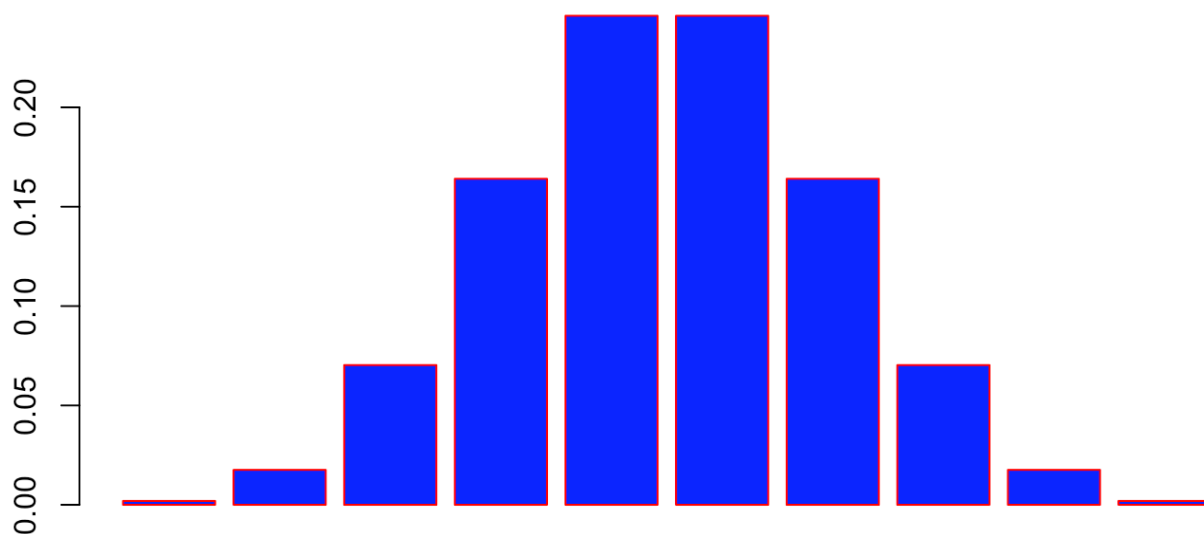
## R实现

```
n = 9; p = 0.1; k = 0
dbinom(x=k,size=n,prob=p)
```

```
##[1] 0.3874205
```

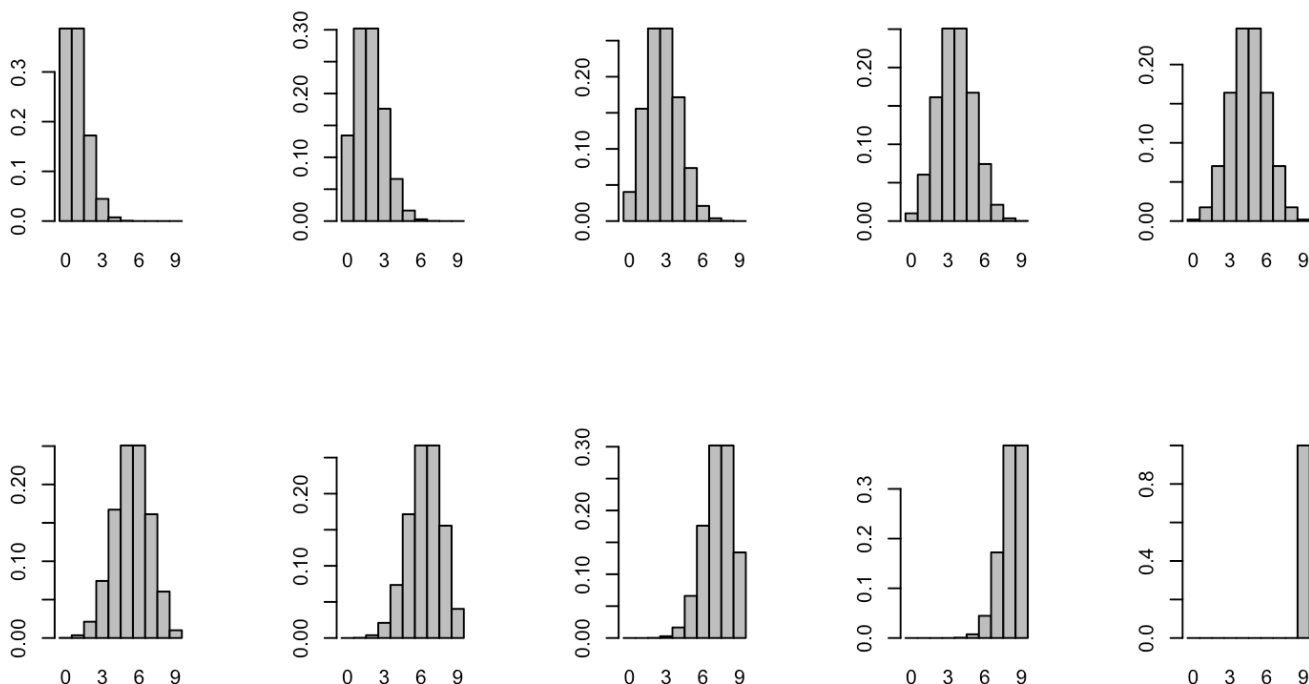
### 画图

```
### https://blog.csdn.net/zfy0910/article/details/120631214
x = seq(0,9,by = 1) # 创建一个0到9的数组间隔为1
y = dbinom(x,9,0.1) #分别求出实现实验成功从0到50次的概率
plot(x,y) #画出二项分布(p=0.1)的图像
z = dbinom(x,9,0.5) #p=0.5; 概率越接近0.5, 图像越对称
plot(x,z)
barplot(z,col="blue",border="red") #实现柱状图
```



多图表示

```
### https://oomake.com/question/5068173
## 方法1:mapply函数
graph <- function(n,p){
  x <- dbinom(0:n,size=n,prob=p)
  barplot(x,names.arg=0:n,
          main=sprintf(paste('bin. dist. ',n,p,sep=':')))
}
op = par(mfcol=c(2,5))
mapply(graph,9,seq(0.1,1,0.1))
par(op)
## 方法2:for循环
par(mfrow=c(2, 5))
for(p in seq(0.1, 1, len=10))
{
  x <- dbinom(0:9, size=9, p=p)
  barplot(x, names.arg=0:9, space=0)
}
```



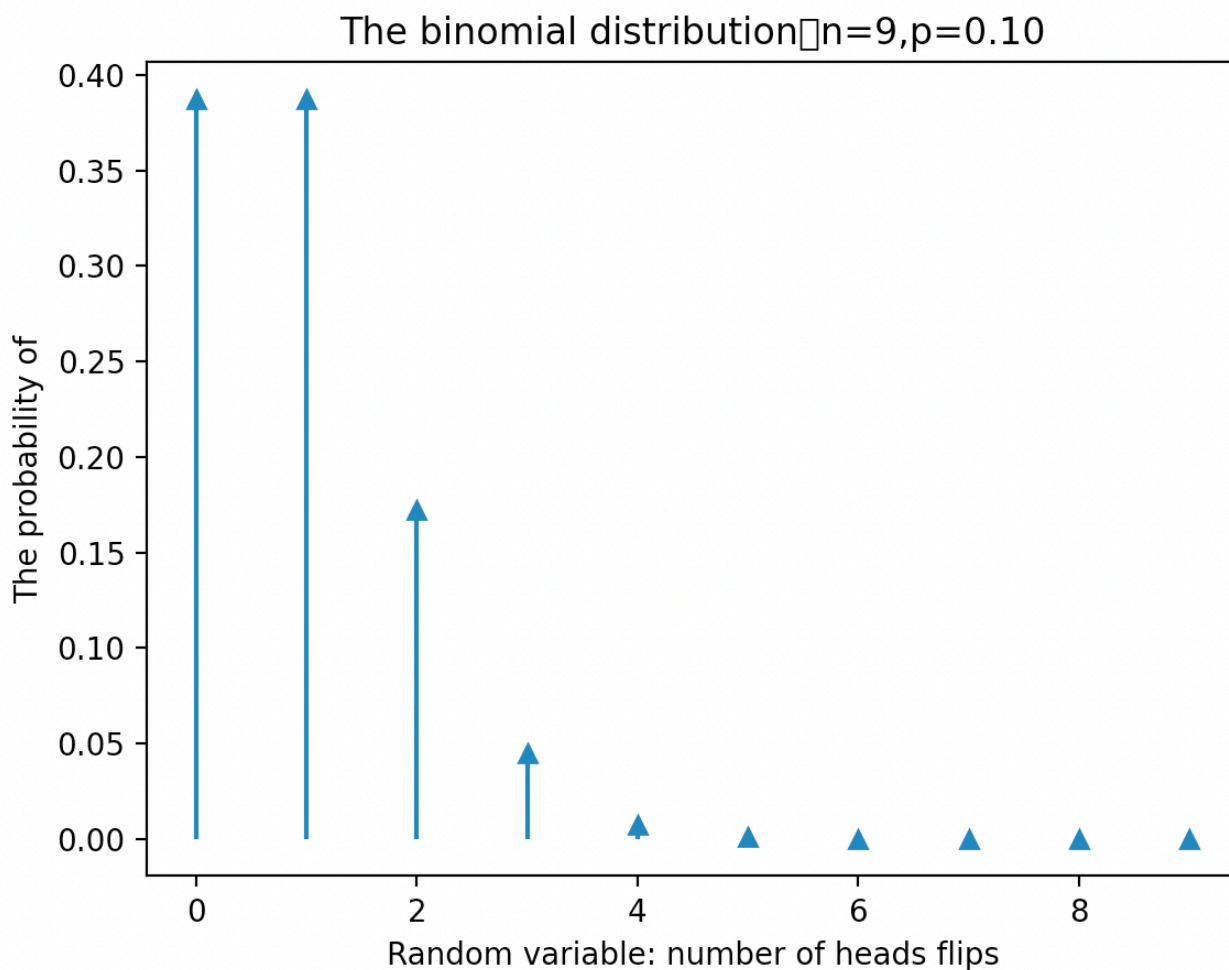
## Python实现

概率精确值

```

### https://docs.scipy.org/doc/scipy/reference/generated/scipy.stats.binom.html
# from scipy import stats
import scipy.stats as stats #导入统计计算包的统计模块
import numpy as np #数组包
import matplotlib.pyplot as plt #绘图包
### arange用于生成一个等差数组, arange(start, stop, step)
n = 9 # 做某件事情的次数
p = 0.1 # 做某件事情成功的概率
X = np.arange(0, n+1, 1)
X
### 求对应分布的概率:概率质量函数 (PMF)
### 它返回一个列表, 列表中每个元素表示随机变量中对应值的概率
pList = stats.binom.pmf(X, n, p)
pList[0] # x=0(即所有勘探失败)的概率
### 绘图
plt.rcParams['font.sans-serif']=['SimHei'] #python 绘图中文显示
plt.plot(X, pList, marker='^', linestyle='None')
plt.vlines(X, 0, pList)
plt.xlabel('随机变量: 抛硬币正面朝上次数') #x轴文本
plt.ylabel('概率') #y轴文本
plt.title('二项分布: n=%i, p=%.2f' % (n, p)) #标题
plt.show() #显示图形
### 更多绘图
# https://blog.csdn.net/weixin\_42695959/article/details/84037126

```



随机数模拟值

```
import numpy as np
n, p = 9, 0.1
a = sum((np.random.binomial(n, p, size=30000)==0))/30000
print(a)
```

```
##0.3853666666666667
```

## 参考

- 何书元. 概率论[M]. 北京: 北京大学出版社, 2006.
- [二项分布和多项分布](#)
- [概率分布的介绍和使用](#)
- [Python统计分析\(3\)——几种重要的概率分布: 二项式分布](#)

## 超几何分布

超几何分布是统计学上一种离散概率分布。它描述了从有限N个物件（其中包含M个指定种类的物件）中抽出n个物件，成功抽出该指定种类的物件的次数（不放回）。称为超几何分布，是因为其形式与“超几何函数”的级数展式的系数有关。

超几何分布中的参数是 $M, N, n$ ，上述超几何分布记作 $X \sim H(n, M, N)$ 。

概率一般记作： $\frac{C_m^a C_n^b}{C_{n+m}^{a+b}}$

## 例

在一个口袋中装有30个球，其中有10个红球，其余为白球，这些球除颜色外完全相同。游戏者一次从中摸出5个球。恰好摸到4个红球就中一等奖，那么获一等奖的概率是多少？

我们手动用公式计算得： $\frac{C_{10}^4 C_{20}^1}{C_{30}^5} = 0.029$

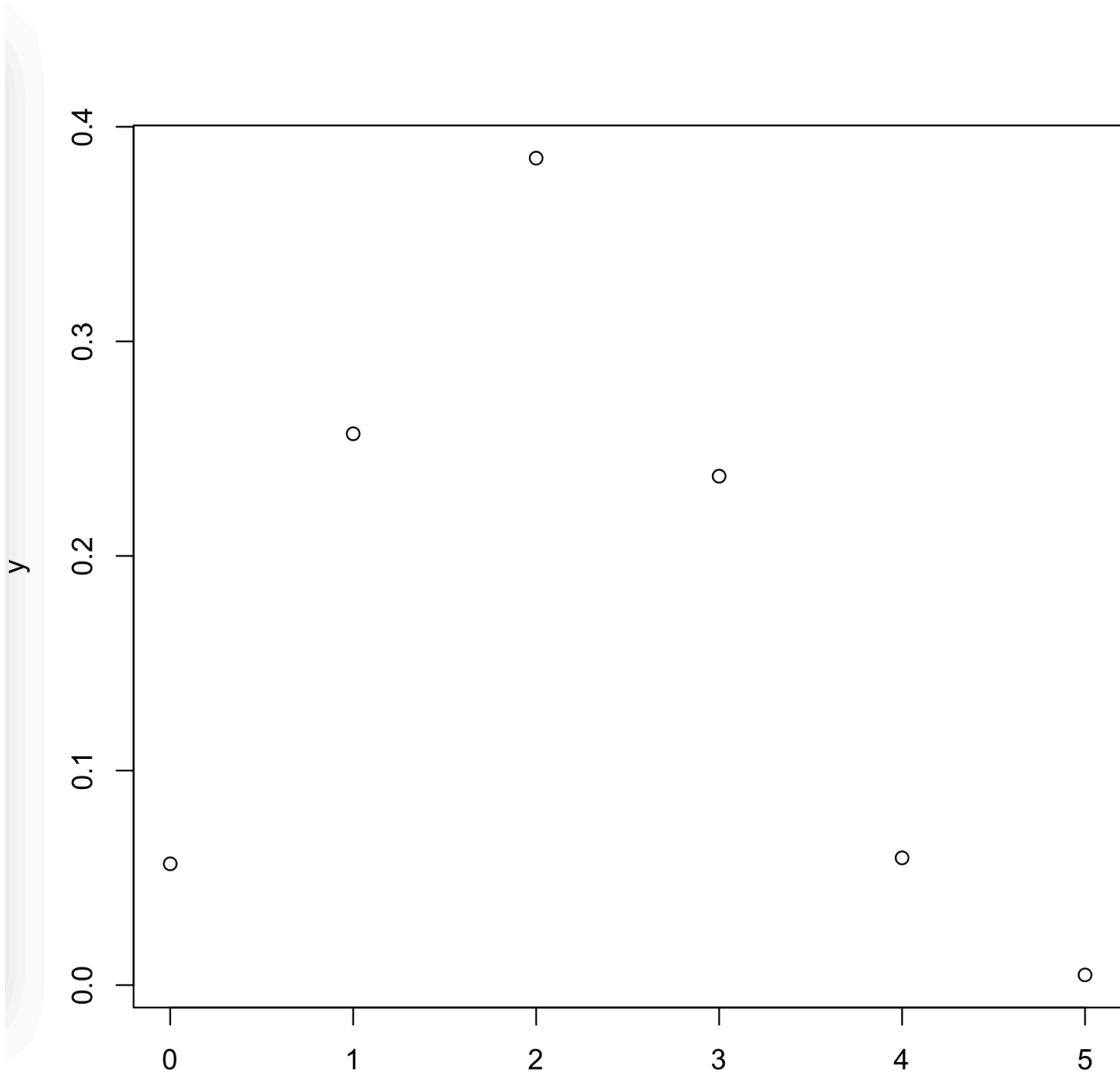
## R实现

```
n<-10      #有十个红球
m<-30-n    #白球的数量
t<-5       #摸球数量
dhyper(4, n, m, t)
```

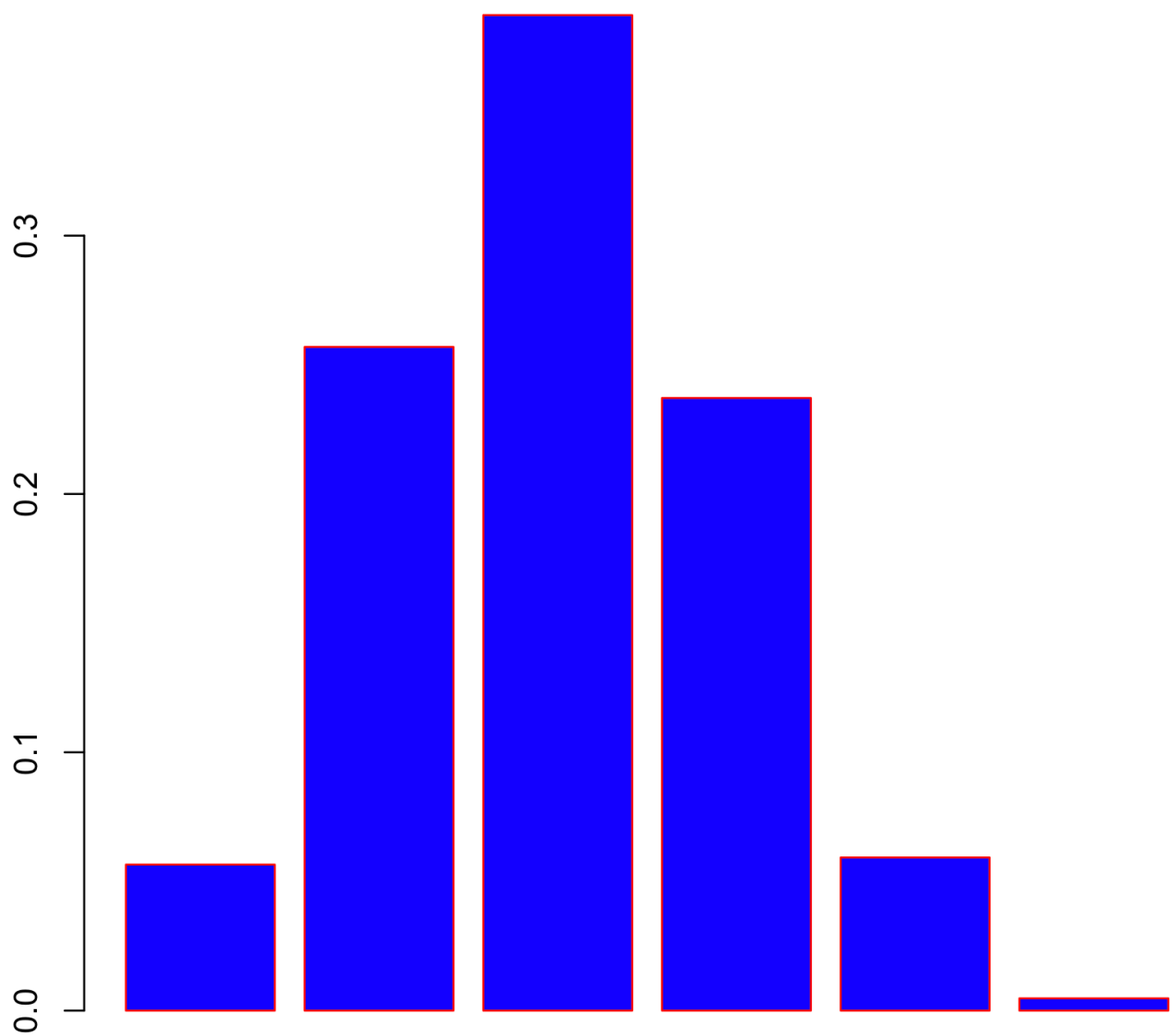
```
##[1] 0.02947244
```

## 画图

```
x = seq(0,5,by = 1) # 创建一个 0 到 5 的数组间隔为 1
y = dhyper(x,5,20,10) # 分别求出实现实验成功从 0 到 5 次的概率
plot(x,y) # 画出超几何分布的图像
```



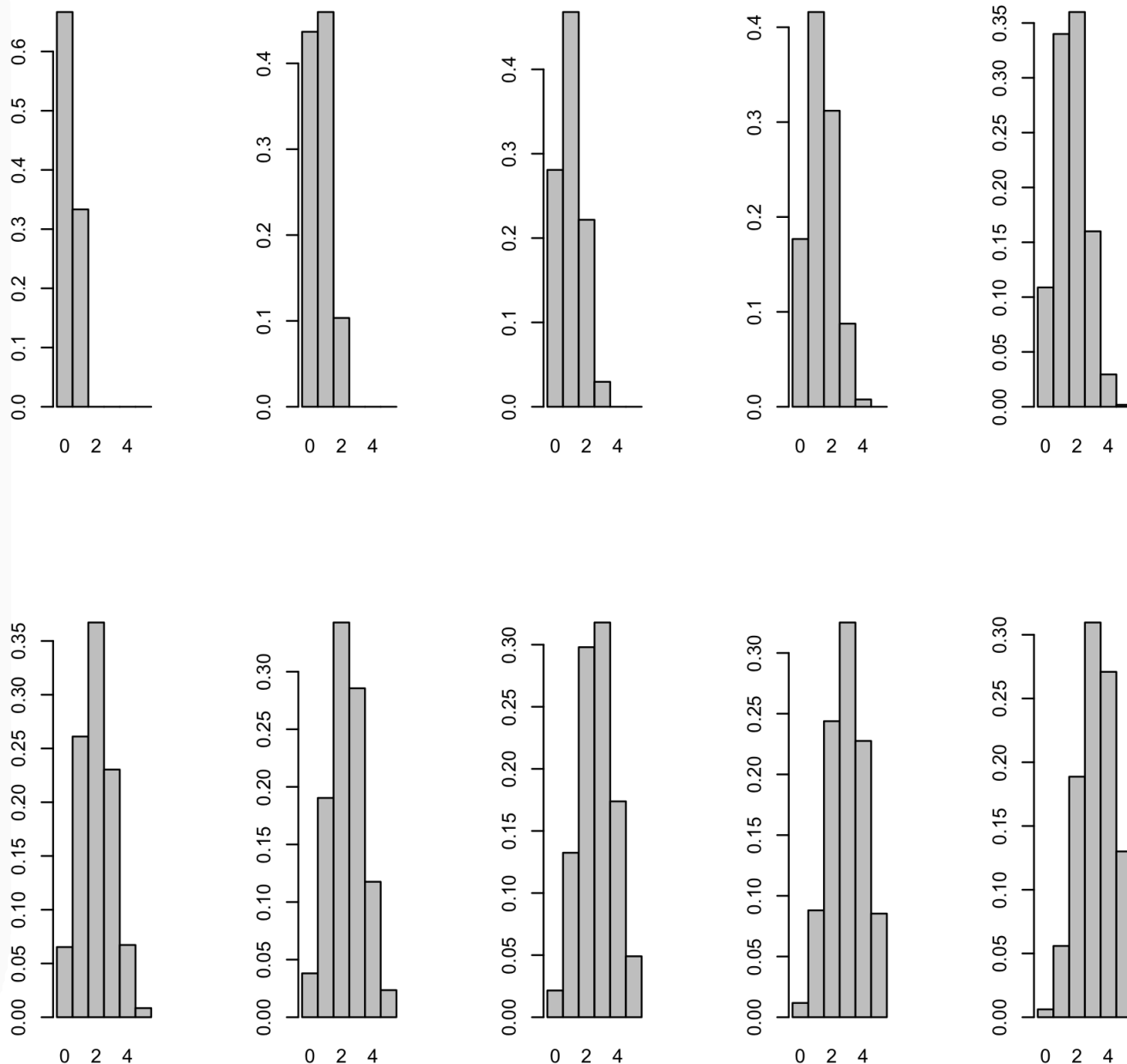
```
barplot(y,col="blue",border="red") # 柱状图
```



多图表示：

```
par(mfrow=c(2, 5))  
for(p in seq(1,10,by=1))  
{x <- dhyper(0:5,p,30-p,10)  
  barplot(x, names.arg=0:5,space=0)}
```



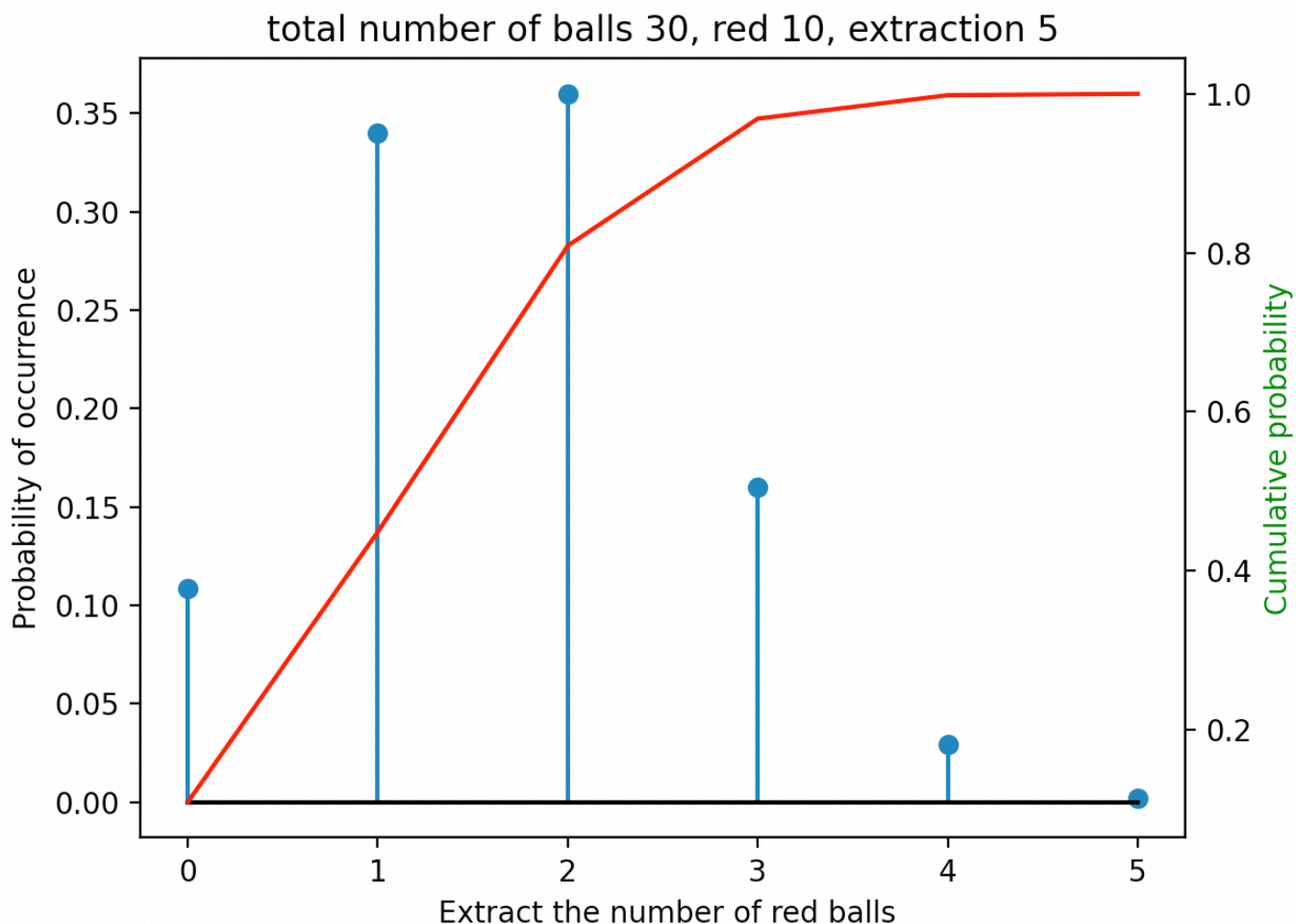


## Python实现

```
import scipy.stats as stats
import numpy as np
import matplotlib.pyplot as plt

N = 30
M = 10
n = 5
po=stats.hypergeom(N,M,n)
x=np.arange(n+1)
px=po.pmf(x)
fig = plt.figure()
ax = plt.gca()
line1 = ax.stem(x,px,basefmt='k',label='The probability of picking a red ball')
ax.set_xlabel(' Extract the number of red balls')
```

```
ax.set_ylabel('Probability of occurrence')
ax.set_title('total number of balls 30, red 10, extraction 5')
ax2=plt.twinx()
y=po.cdf(x)
line2 = ax2.plot(x,y,'r',label='Cumulative probability')
ax2.set_ylabel('Cumulative probability',color='g')
plt.show()
```



随机数模拟值

```
import numpy as np
n, p = 9, 0.1
s = np.random.hypergeometric(10,20,5,size=1000000)
print(s)
```

```
##[3 2 0 ... 2 1 1]
```

## 参考

- 盛骤, 谢式千, 潘承毅. 概率论与数理统计[M]. 第4版. 北京: 高等教育出版社, 2008.
- 王学民. 应用多元统计分析[M]. 第5版. 上海: 上海财经大学出版社, 2017.

## 多项分布

如果一个随机向量  $X = (X_1, X_2, \dots, X_N)$  满足下列条件:

(1)  $X_i \geq 0$  ( $1 \leq i \leq n$ ), 且  $X_1 + X_2 + \dots + X_n = N$ ;

(2) 设  $m_1, m_2, \dots, m_n$  为任意非负整数, 且  $m_1 + m_2 + \dots + m_n = N$ , 则事件

$X_1 = m_1; X_2 = m_2, \dots, X_n = m_n$  的概率为

$$P\{X_1 = m_1, X_2 = m_2, \dots, X_n = m_n\} = \frac{N!}{m_1!m_2!\dots m_n!} p_1^{m_1} p_2^{m_2} \dots p_n^{m_n}$$

其中,  $p_i \geq 0$  ( $1 \leq i \leq n$ ),  $p_1 + p_2 + \dots + p_n = 1$  则称随机向量  $X = (X_1, X_2, \dots, X_n)$  服从多项分布, 记作  $X \sim PN(N; p_1, p_2, \dots, p_n)$ 。

## 例

西班牙足球甲级联赛(简称“西甲”)已经成为世界上最高水平的足球联赛之一, 吸引了世界球迷的目光, 巴萨与皇马的德比之战更是令世界瞩目。截止目前, 巴萨在历史上和皇马交手178次, 拿到了71胜35平72负的战绩。如果接下来巴萨与皇马将进行6场比赛, 求巴萨3胜1平2负的概率是多少?

我们手动用公式计算得:

$$P\{X_1 = m_1, X_2 = m_2, \dots, X_n = m_n\} = \frac{N!}{m_1!m_2!\dots m_n!} p_1^{m_1} p_2^{m_2} \dots p_n^{m_n}$$

$$P\{X_1 = \frac{71}{178}, X_2 = \frac{35}{178}, X_3 = \frac{72}{178}\} = \frac{6!}{3!1!2!} p_1^{m_1} p_2^{m_2} p_3^{m_3}$$

其中:  $p_1 = \frac{71}{178}, p_2 = \frac{35}{178}, p_3 = \frac{72}{178}, m_1 = 3, m_2 = 1, m_3 = 2$   
得到概率为: 0.1225009

## R实现

```
multi <- function(p, x) {  
  n <- sum(x)    ##试验次数  
  
  f1 <- prod(p^x)  
  f2 <- prod(factorial(x))  
  factorial(n)*f1/f2    ##返回值  
}  
p <- c(71, 35, 72)/178  
x <- c(3, 1, 2)  
multi(p, x)
```

```
##[1] 0.1225009
```

R模拟:

```

a <- c('胜', '平', '负')
p <- c(71, 35, 72)/178 ##巴萨胜、平、负各概率值
n <- 10000 ##比赛场数

set.seed(18)
result <- sample(a, n, replace=TRUE, prob=p) ##replace=TRUE表示可重复抽样

head(result, 20) ##前20场比赛的结果
table(result) ## 计数
result
n*p[1] ##巴萨获胜场数的理论值
n*p[2] ##巴萨打平场数的理论值
n*p[3] ##巴萨输皇马场数的理论值

```

```

> a <- c('胜', '平', '负')
> p <- c(71, 35, 72)/178 ##巴萨胜、平、负各概率值
> n <- 10000 ##比赛场数
>
> set.seed(18)
> result <- sample(a, n, replace=TRUE, prob=p) ##replace=TRUE表示可重复抽样
>
> head(result, 20) ##前20场比赛的结果
[1] "平" "胜" "平" "负" "负" "胜" "负" "胜" "负" "负" "胜" "胜" "负" "负" "平" "负"
[17] "胜" "负" "平" "负"
> table(result) ## 计数
result
  负   平   胜
4022 1958 4020
> result
[1] "平" "胜" "平" "负" "负" "胜" "负" "胜" "负" "负" "胜" "胜" "负" "负" "平" "负"
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[609] "平" "平" "胜" "负" "负" "胜" "平" "胜" "胜" "负" "平" "负" "胜" "胜" "胜" "负"
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[673] "负" "负" "胜" "胜" "胜" "负" "胜" "胜" "负" "平" "负" "平" "胜" "平" "平" "胜"
[689] "平" "胜" "平" "胜" "胜" "负" "胜" "胜" "胜" "胜" "胜" "胜" "胜" "平" "负" "平"
[705] "胜" "胜" "胜" "负" "胜" "平" "平" "负" "负" "胜" "负" "胜" "胜" "平" "胜" "负"
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[737] "平" "负" "胜" "胜" "负" "负" "胜" "胜" "负" "负" "负" "胜" "负" "平" "胜" "平"
[753] "胜" "胜" "胜" "胜" "负" "平" "平" "胜" "胜" "胜" "平" "胜" "胜" "负" "负" "胜"
[769] "平" "胜" "负" "胜" "胜" "平" "负" "胜" "负" "负" "负" "胜" "胜" "负" "负" "负"
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[801] "负" "平" "负" "负" "负" "胜" "负" "负" "胜" "胜" "负" "胜" "胜" "负" "负" "负"
[817] "负" "胜" "胜" "负" "负" "负" "负" "胜" "负" "平" "胜" "负" "负" "负" "负" "胜"
[833] "平" "胜" "负" "负" "平" "负" "胜" "负" "负" "胜" "负" "胜" "平" "胜" "负" "负"
[849] "负" "平" "胜" "平" "负" "负" "负" "负" "胜" "胜" "平" "胜" "平" "胜" "胜" "平"
[865] "胜" "负" "平" "负" "平" "负" "胜" "平" "胜" "胜" "负" "胜" "胜" "平" "胜" "胜"
[881] "胜" "负" "负" "负" "负" "负" "平" "平" "平" "胜" "平" "胜" "平" "负" "胜" "负"
[897] "负" "胜" "负" "负" "平" "胜" "负" "负" "负" "负" "负" "平" "胜" "平" "负" "负"
[913] "平" "胜" "平" "胜" "负" "平" "负" "平" "平" "负" "负" "负" "负" "胜" "负" "负"
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[945] "平" "胜" "负" "胜" "负" "平" "负" "负" "胜" "负" "平" "平" "胜" "负" "负" "胜"
[961] "负" "胜" "负" "平" "胜" "胜" "负" "负" "平" "胜" "负" "胜" "胜" "胜" "负" "胜"
[977] "平" "平" "胜" "平" "胜" "平" "胜" "胜" "负" "平" "胜" "负" "负" "负" "平" "胜"
[993] "平" "负" "胜" "胜" "负" "胜" "平" "平"

```

```
[ reached getOption("max.print") -- omitted 9000 entries ]
```

```
> n*p[1] ##巴萨获胜场数的理论值
```

```
[1] 3988.764
```

```
> n*p[2] ##巴萨打平场数的理论值
```

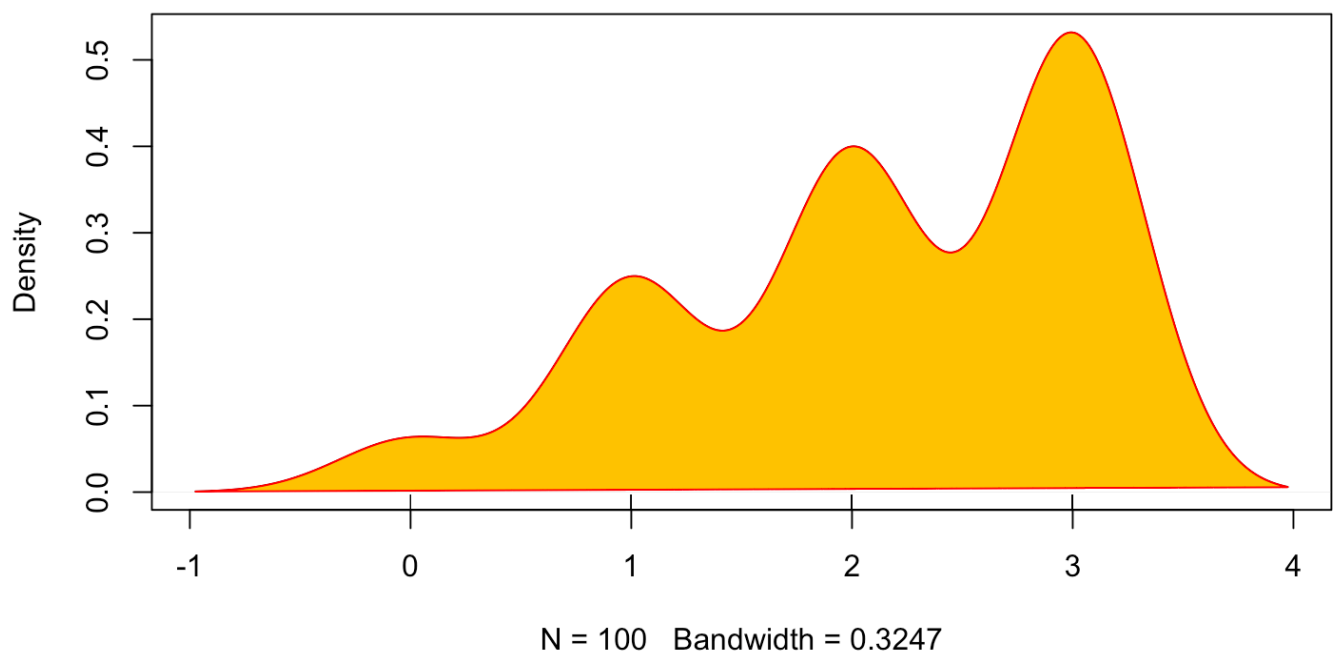
```
[1] 1966.292
```

```
> n*p[3] ##巴萨输皇马场数的理论值  
[1] 4044.944
```

画图

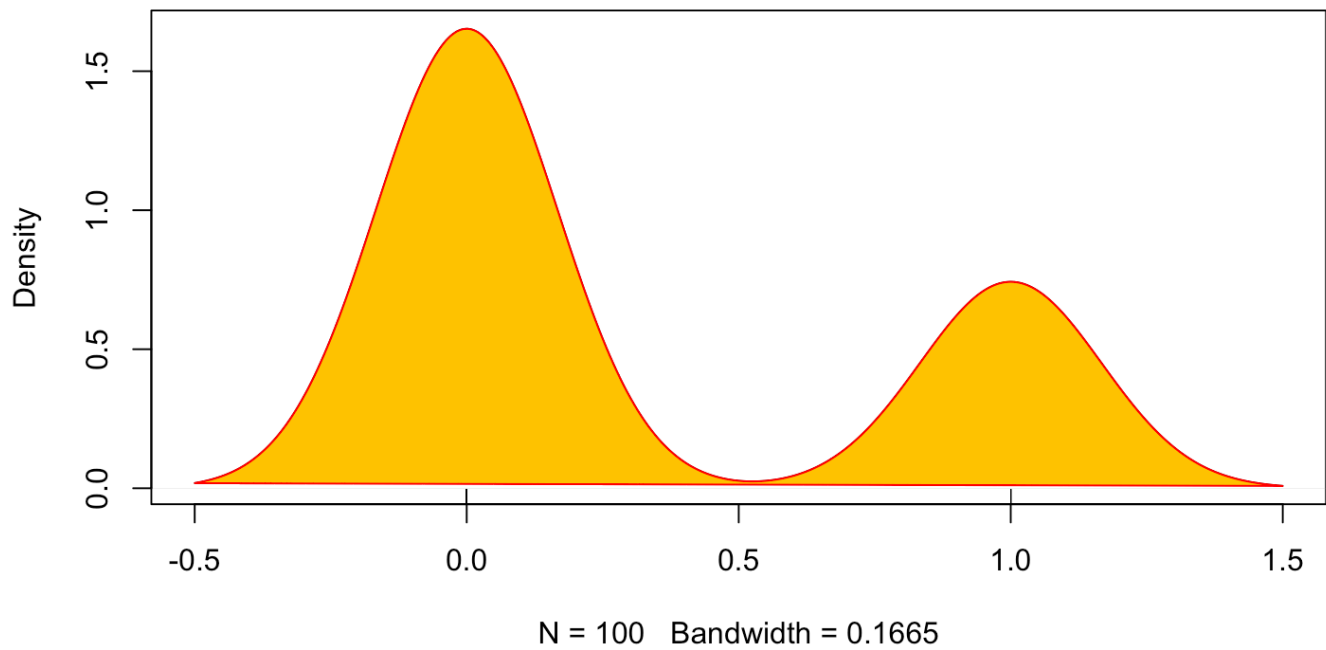
```
x1<-round(rhyper(nn=100,m=71,n=35,k=3));x1  
plot(density(x1),main = "Hypergeometric distribution kernel density diagram")  
polygon(density(x1),col=7,border = "red")  
rug(x1,col="black")
```

### Hypergeometric distribution kernel density diagram



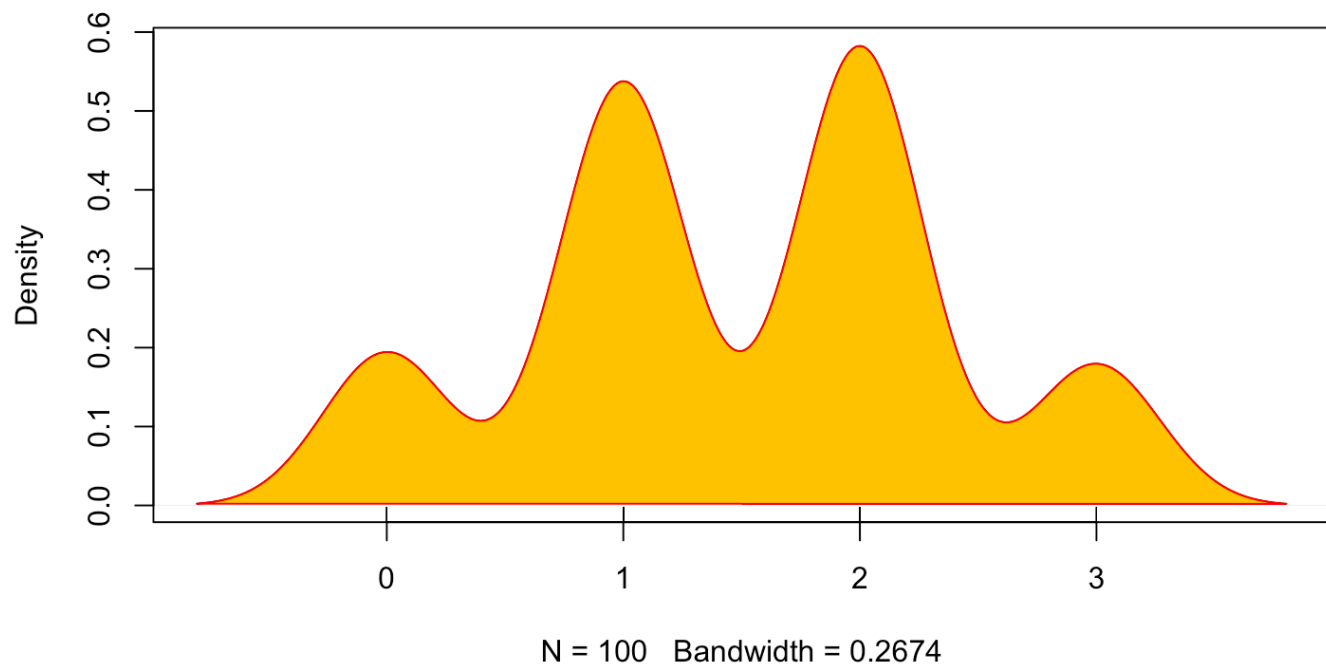
```
x2<-round(rhyper(nn=100,m=35,n=72,k=1));x2  
plot(density(x2),main = "Hypergeometric distribution kernel density diagram")  
polygon(density(x2),col=7,border = "red")  
rug(x2,col="black")
```

### Hypergeometric distribution kernel density diagram



```
x3<-round(rhyper(nn=100,m=71,n=72,k=3));x3
plot(density(x3),main = "Hypergeometric distribution kernel density diagram")
polygon(density(x3),col=7,border = "red")
rug(x3,col="black")
```

### Hypergeometric distribution kernel density diagram



**Python实现**

```

from functools import reduce
a = reduce(lambda x,y:x*y,range(1,6+1))
b = reduce(lambda x,y:x*y,range(1,3+1))
c = reduce(lambda x,y:x*y,range(1,1+1))
d = reduce(lambda x,y:x*y,range(1,2+1))
p = ((a)/(b*c*d))*((71/178)**3)*(35/178)*((72/178)**2)
print(p)

```

```
##0.1225009114171614
```

python模拟

```

import numpy as np
from numpy import random
for i in range(1):
    a = np.random.multinomial(10000, [71/178,35/178,72/178], size=1)
    print(a)

```

```
##[[3961 1942 4097]]
```

## 参考

- 王学民. 应用多元统计分析[M]. 第5版. 上海: 上海财经大学出版社, 2017.  
-<https://zhuanlan.zhihu.com/p/52481385>

## 多元超几何分布

多元超几何分布亦称阶乘多项分布。超几何分布的多元推广。

有一批产品共 $N$ 件，其中一等品 $M_1$ 件，二等品 $M_2$ 件，从此批产品中任意取 $n$ 件，求恰取出 $k_1$ 件一等品，求 $k_2$ 件二等品的概率。

设抽出的 $n$ 件产品，有一等品的 $X$ 件，二等品 $Y$ 件，从 $N$ 件产品中取出 $n$ 件，共有取法 $C_N^n$ 种。 $(X = k_1, Y = k_2)$ 这件事相当于 $M_1$ 件一等品中取 $k_1$ 件，从 $M_2$ 件二等品中取 $k_2$ ，再从余 $N - M_1 - M_2$ 件产品中取出 $n - k_1 - k_2$ 件。

由乘法原理得： $P(X = k_1, Y = k_2) = \frac{C_{M_1}^{k_1} C_{M_2}^{k_2} C_{N-M_1-M_2}^{n-k_1-k_2}}{C_N^n}$

$k_1 = 0, 1, 2, \dots, \min\{n, M_1\}, k_2 = 0, 1, 2, \dots, \min\{n, M_2\}$

## 例

有一批产品100件，其中一等品70件，二等品30件，从此批产品中任意取10件，求恰好取出7件一等品，3件二等品的概率？

我们手动用公式计算得：

设：有一等品的 $X$ 件，二等品 $Y$ 件。

$P(X = 7, Y = 3) = \frac{C_{70}^7 C_{30}^3 C_{100-70-30}^{10-7-3}}{C_{100}^{10}} = 0.2811633943$



## R实现

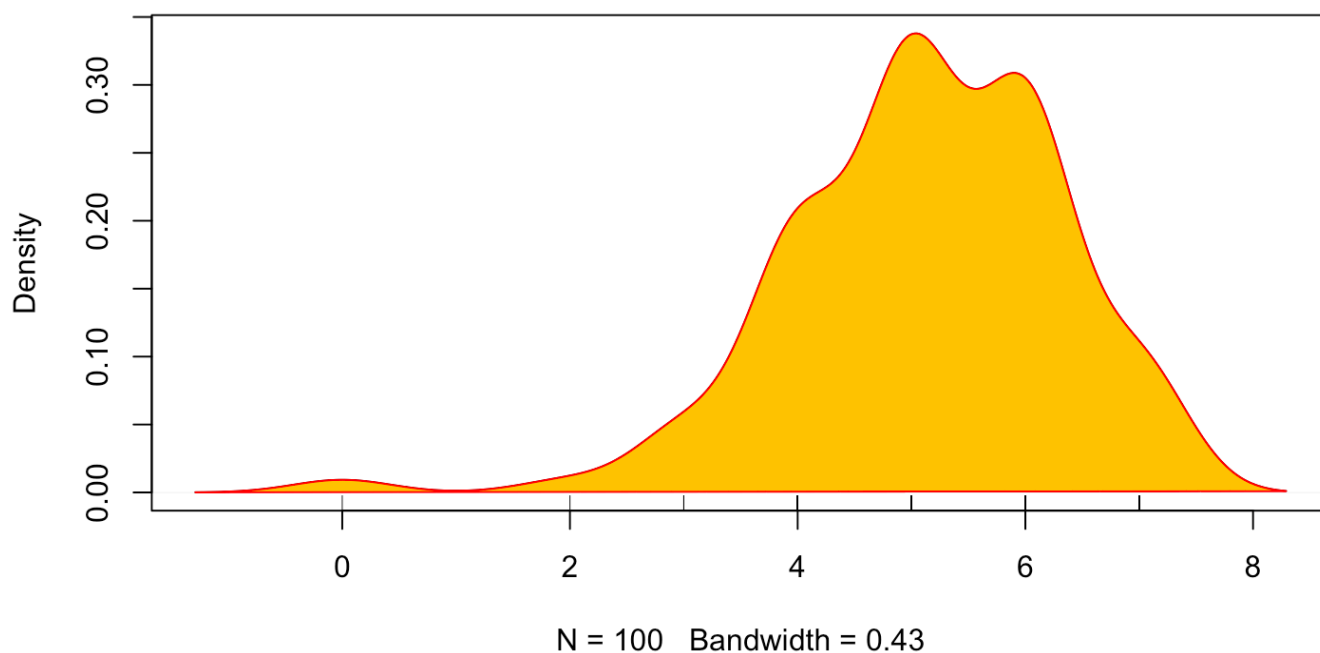
```
choose(70,7)*choose(30,3)/choose(100,10)
```

```
##[1] 0.2811634
```

画图

```
x1<-round(rhyper(nn=100,m=70,n=30,k=7));x1  
plot(density(x1),main = "Hypergeometric distribution kernel density diagram")  
polygon(density(x1),col=7,border = "red")  
rug(x1,col="black")
```

Hypergeometric distribution kernel density diagram



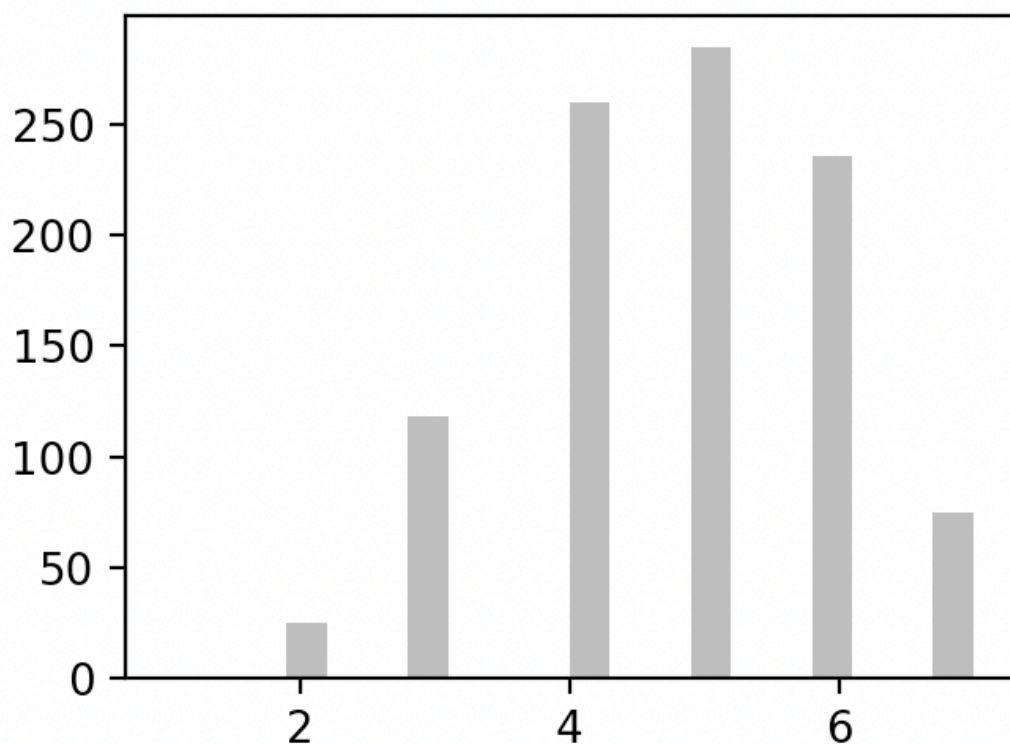
## Python实现

```
from scipy.stats import multivariate_hypermgeom  
a = multivariate_hypermgeom.pmf(x =[7,3], m =[70,30],n = 10)  
#在70个一等品中取7个，30个二等品中取3个，总样本为10  
print(a)
```

```
##0.2811633943025375
```

python模拟

```
import numpy as np
import matplotlib.pyplot as plt
s = np.random.hypergeometric(70,30,7,size=1000)
p = 1- sum(s>=4)/1000.
fig = plt.figure(figsize=(8,6))
a1 = fig.add_subplot(2,2,1)
a1.hist(s ,bins=20,color='k',alpha=0.3)
plt.show()
```



随机数模拟

```
import numpy as np
n, p = 9, 0.1
s = np.random.hypergeometric(70,30,7,size=1000000)
print(s)
```

```
##[3 5 5 ... 6 6 6]
```

参考

- 王学民. 应用多元统计分析[M]. 第5版. 上海: 上海财经大学出版社, 2017.