

MW2

yp2201 @ nyu.edu

1. Given  $\hat{a}$  a geometric random variable with parameter  $d$ ,

geometric distribution is:  $P_{\hat{a}}(a) = (1-d)^{a-1} \cdot d^1$

if we assume  $a_1 = \hat{a}$  equals  $a$  for  $a=1,2,3,\dots$ ,

and  $a_2 = \hat{a}$  equals  $a$  for  $a > 5$ ,

the probability should be  $P_{\hat{a}}(a_1) / P_{\hat{a}}(a_2) = \frac{P_{\hat{a}}(a_1 \cap a_2)}{P_{\hat{a}}(a_2)}$

(f)  $a=1,2,3,4,5$  then  $P_{\hat{a}}(a_1 \cap a_2) = 0$ ,

Since there will be no intersection of value  $a$ .

if  $a > 5$ , then  $P_{\hat{a}}(a_1 \cap a_2) = P_{\hat{a}}(a_1)$

Since every  $a_1$  will be an intersection of  $a_2$

$$P_{\hat{a}}(a_1) = (1-d)^{(a_1-1)} \cdot d^1.$$

$$P_{\hat{a}}(a_2) = P_{\hat{a}}(a > 5) = 1 - (P_{\hat{a}}(1) + P_{\hat{a}}(2) + P_{\hat{a}}(3) + P_{\hat{a}}(4) + P_{\hat{a}}(5))$$

$$= 1 - (d + (1-d)d + (1-d)^2d + (1-d)^3d + (1-d)^4d)$$

$$= (1-d) - (1-d)d - (1-d)^2d - (1-d)^3d - (1-d)^4d$$

$$= (1-d)(1-d - (1-d)d - (1-d)^2d - (1-d)^3d)$$

$$= (1-d)^2(1-d - (1-d)d - (1-d)^2d)$$

$$= (1-d)^3(1-d - (1-d)d)$$

$$= (1-d)^4(1-d)$$

$$= (1-d)^5$$

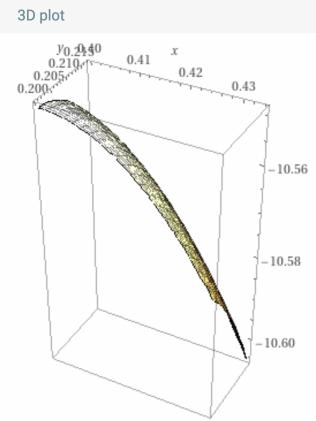
if  $a \leq 5$ ,  $P_{\hat{a}}(a_1) / P_{\hat{a}}(a_2) = \frac{P_{\hat{a}}(a_1 \cap a_2)}{P_{\hat{a}}(a_2)} = 0$

$a > 5$ ,  $P_{\hat{a}}(a_1) / P_{\hat{a}}(a_2) = \frac{P_{\hat{a}}(a_1 \cap a_2)}{P_{\hat{a}}(a_2)} = \frac{P_{\hat{a}}(a_1)}{P_{\hat{a}}(a_2)} = \frac{(1-d)^{(a_1-1)} \cdot d^1}{(1-d)^5} = \boxed{\frac{(1-d)^{(a_1-6)} \cdot d^1}{(1-d)^5}}$

$$Q. 2. (a) P(\theta, a) = \frac{10!}{4!4!2!} \theta^4 a^2 (1-\theta-a)^4$$

$$\log P(\theta, a) = \log \frac{10!}{4!4!2!} + 4 \log(\theta) + 2 \log(a) + 4 \log(1-\theta-a)$$

(Plot 7)



(b) By taking the derivative of both  $\theta$  and  $a$ , we can calculate  $\theta, a$ .

$$1) \frac{d}{d\theta} \log P(\theta, a) = 0 + \frac{4}{\theta} + 0 + \frac{4 \times (-1)}{(1-\theta-a)} = 0$$

$$\Rightarrow \frac{4}{\theta} = \frac{4}{(1-\theta-a)}, \quad \theta = (1-\theta-a) \quad : \quad \theta = \frac{1-a}{2}$$

$$2) \frac{d}{da} \log P(\theta, a) = 0 + 0 + \frac{2}{a} + \frac{4(-1)}{(1-\theta-a)} = 0, \quad 2a = 1-\theta-a \quad : \quad \theta = 1-3a$$

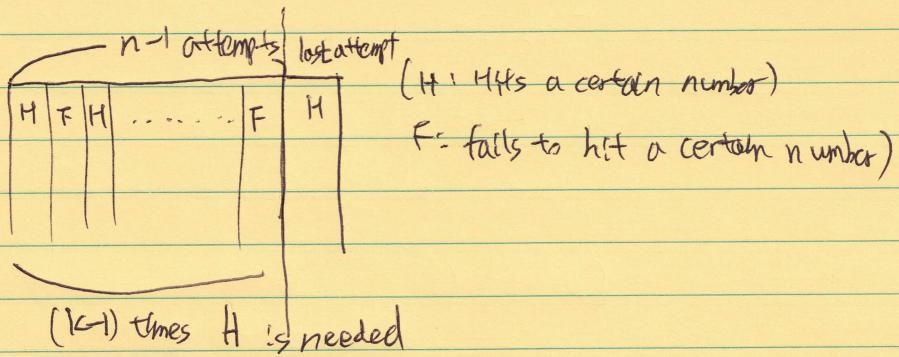
Since  $\theta = \frac{1-a}{2} = 1-3a$ ,  $1-a=2-6a$ ,  $5a=1$ ,  $[a=\frac{1}{5}]$ ,  $\theta = \frac{1-\frac{1}{5}}{2} = \frac{2}{5}$

(c)  $P(\text{Garry wins}) = \frac{4}{10} = \frac{2}{5}$  from 10 chess games.

$P(\text{Anish wins}) = \frac{2}{10} = \frac{1}{5}$  This result is same as (b),

$P(\text{draw}) = \frac{4}{10} = \frac{2}{5}$ . which makes this nonparametric model same as parametric model

3. If a player needs total  $n$  attempts and hits a certain number  $k$  times, last attempt should be hitting a certain number.



So, for  $(n-1)$  attempts, we have  $(k-1)$  Hits and  $(n-1)-(k-1)$  none hits in binomial distribution.

Assuming the probability of success in each attempt as  $\theta$ , and number of required attempts  $k$ , pmf should be as below

$$\text{pmf) } \frac{n-1 C_{k-1} \cdot (\theta)^{k-1} \cdot (1-\theta)^{(n-1)-(k-1)}}{p(\text{last attempt} = \theta)}$$

$$= n-1 C_{k-1} (\theta)^k \cdot (1-\theta)^{n-k}$$

```
In [13]: # 4. (Call center) Complete the code in the iPython notebook:
def empirical_pmf(months, max_count, time_ini, time_end, verbose):
    # initializing count result
    result = np.zeros(max_count)

    # loop months and count for each months
    for month in months:
        # counts value for certain month
        counts = compute_counts(time_ini, time_end, month, verbose)

        # Feb, Apr, June, Sep, Nov has days less than 31.
        # slice counts to exact days
        if month_number[month] == 2:
            counts = counts[:28]
        elif month_number[month] == 4 or 6 or 9 or 11:
            counts = counts[:30]

        # for every day, if given day is not weekend, then add to count
        # add counts to certain result index
        for i in range(len(counts)):
            if is_weekend(month_number[month], i+1) == False:
                result[int(counts[i])] += 1

    # get final pmf
    pmf = result / sum(result)

    # return its value
    return pmf
```

```
# import math to use factorial function
import math

def poisson_model(months, max_count, time_ini, time_end, verbose):

    # initialize two ndarrays with max_count, also initialize total calls and number of days
    pmf = np.zeros(max_count)
    result = np.zeros(max_count)
    totalCalls = 0
    numDays = 0

    # loop months and count for each months
    for month in months:
        # counts value for certain month
        counts = compute_counts(time_ini, time_end, month, verbose)

        # Feb, Apr, June, Sep, Nov has days less than 31.
        # slice counts to exact days
        if month_number[month] == 2:
            counts = counts[:28]
        elif month_number[month] == 4 or 6 or 9 or 11:
            counts = counts[:30]

        # for every day, if given day is not weekend, then count
        # add total counts to totalcalls and add 1 for each success if statement
        for i in range(len(counts)):
            if is_weekend(month_number[month], i+1) == False:
                totalCalls += counts[i]
                numDays += 1

    # lambda should be mean value, so divide total calls by number of days
    param = totalCalls / numDays

    # loop x from 0 to length of pmf, and append possion values to pmf[x]
    for x in range(len(pmf)):
        pmf[x] = (np.exp(-param) * (param ** x)) / math.factorial(x)

    # return pmf
    return pmf
```