

Assignment 1.

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Due Oct 8

1. (a)

$$\Omega = \frac{N!}{N_h!(N - N_h)!}$$

$$\log \Omega = \log(N!) - \log(N_h!) - \log((N - N_h)!)$$

Considering N_h as a pertubation about $\frac{N}{2}$,

$$\begin{aligned} \log \Omega &= \log(N!) - \log\left(\left(\frac{N}{2} + \epsilon\right)!\right) - \log\left(\left(\frac{N}{2} - \epsilon\right)!\right) \\ &= N \log N - N + \frac{1}{2} \log(2\pi N) \\ &\quad - \left(\frac{N}{2} + \epsilon\right) \log\left(\frac{N}{2} + \epsilon\right) + \left(\frac{N}{2} + \epsilon\right) - \frac{1}{2} \log(2\pi\left(\frac{N}{2} + \epsilon\right)) \\ &\quad - \left(\frac{N}{2} - \epsilon\right) \log\left(\frac{N}{2} - \epsilon\right) + \left(\frac{N}{2} - \epsilon\right) - \frac{1}{2} \log(2\pi\left(\frac{N}{2} - \epsilon\right)) \\ &= N \log N + \frac{1}{2} \log(2\pi N) \\ &\quad - \left(\frac{N}{2} + \epsilon\right) \log\left(\left(\frac{N}{2}\right)\left(1 + \frac{2\epsilon}{N}\right)\right) - \frac{1}{2} \log\left(\left(2\pi\frac{N}{2}\right)\left(1 + \frac{2\epsilon}{N}\right)\right) \\ &\quad - \left(\frac{N}{2} - \epsilon\right) \log\left(\left(\frac{N}{2}\right)\left(1 - \frac{2\epsilon}{N}\right)\right) - \frac{1}{2} \log\left(\left(2\pi\frac{N}{2}\right)\left(1 - \frac{2\epsilon}{N}\right)\right) \\ &= N \log N + \frac{1}{2} \log(2\pi N) - N \log(N/2) - \log(\pi N) \\ &\quad - \left(\frac{N}{2} + \epsilon\right) \log\left(1 + \frac{2\epsilon}{N}\right) - \frac{1}{2} \log\left(1 + \frac{2\epsilon}{N}\right) \\ &\quad - \left(\frac{N}{2} - \epsilon\right) \log\left(1 - \frac{2\epsilon}{N}\right) - \frac{1}{2} \log\left(1 - \frac{2\epsilon}{N}\right) \\ &\approx N \log N + \frac{1}{2} \log(2\pi N) - N \log(N/2) - \log(\pi N) \\ &\quad - \left(\frac{N}{2} + \epsilon\right)\left(\frac{2\epsilon}{N} - \frac{2\epsilon^2}{N^2}\right) - \frac{1}{2}\left(\frac{2\epsilon}{N} - \frac{2\epsilon^2}{N^2}\right) \\ &\quad - \left(\frac{N}{2} - \epsilon\right)\left(-\frac{2\epsilon}{N} - \frac{2\epsilon^2}{N^2}\right) - \frac{1}{2}\left(-\frac{2\epsilon}{N} - \frac{2\epsilon^2}{N^2}\right) \\ &\approx N \log N + \frac{1}{2} \log(2\pi N) - N \log(N/2) - \log(\pi N) - \frac{2\epsilon^2}{N} + \frac{2\epsilon^2}{N^2} \end{aligned}$$

$$\Omega = \frac{2^{N+\frac{1}{2}}}{\sqrt{\pi N}} e^{-\frac{2\epsilon^2}{N} + \frac{2\epsilon^2}{N^2}}$$

(b) Maximum multiplicity is $\frac{2^{N+\frac{1}{2}}}{\sqrt{\pi N}}$. It is less than 2^N as $N \geq 0$.

(c) Width is given by $2\epsilon^*$ where

$$\frac{2\epsilon^{2*}}{N} - \frac{2\epsilon^{2*}}{N^2} = 1$$

$$\epsilon^* = \frac{N}{\sqrt{2N+2}}$$

Thus, the width is $\frac{2N}{\sqrt{2N+2}}$

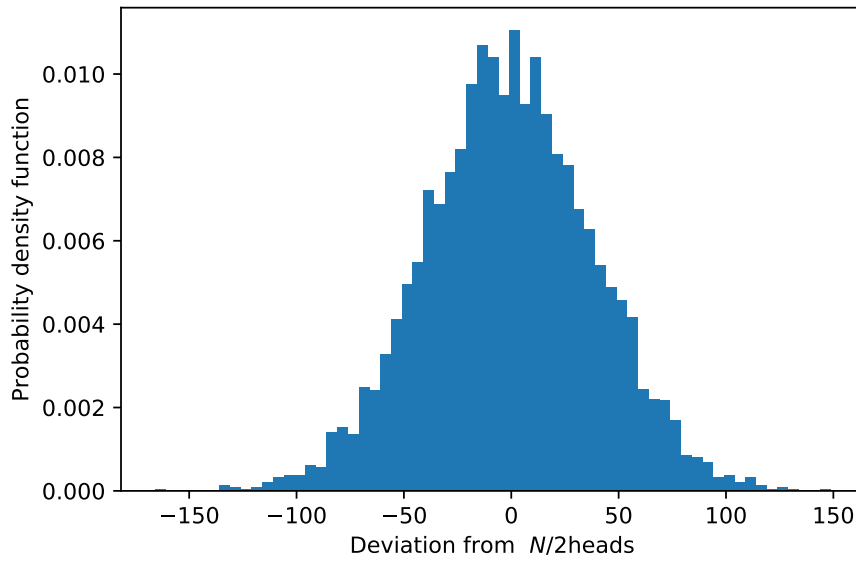


Figure 1: The histogram

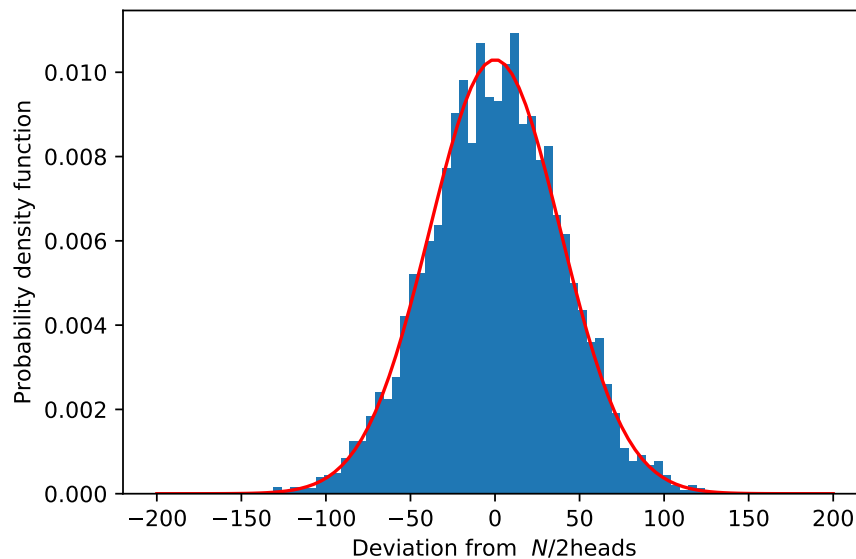


Figure 2: The histogram with the probability density function

2. (a)
- (b) Done in part(a).
- (c) Done in part(a).
- (d) It would be a probability distribution which would be accurate near $\frac{N}{2}$.
- (e) Looking at the Normalization description from the Matlab help page, I think dividing the histogram by the bin width would give us the probability density function.
- (f) This code allows us to obtain the width of distribution analytically.

```
# September 2017 -- Coin toss
import numpy as np
import matplotlib.pyplot as plt
N_trials = 5000;
Sample_sizes = [2000,3000,6000];
# Note that you can run just one or several Sample_sizes.
# For example, replace the line above with Sample_sizes =
# [1000, 2000]
# to just have the two Sample_sizes: 1000 and 2000.
width_experimental = []
width_analytical = []
for k in Sample_sizes:
    N_samples = k
    Total_heads = []
    for i in range(N_trials):
```

```

# Generate [1 x N_samples] vector of
# uniformly distributed random
# integers from 0 or 1.
rs = np.random.randint(2,size =
    N_samples);
# Calculate the total number of
# heads.
Total_heads.append(sum(rs));

BW = 5; # Bin width
data = np.array(Total_heads)-N_samples/2
histdata = plt.hist(data,normed=True, bins=np.arange
    (min(data), max(data) + int(BW), int(BW)))
x = np.linspace(-200, 200,100) # returns a row
    vector of 100 evenly spaced points between -200
    and 200
POmega = (1/np.sqrt(np.pi*N_samples/2))*np.exp(-2*(x
    **2)/N_samples+ -2*(x**2)/N_samples**2);
plt.plot (x, POmega, 'r-')# 'LineWidth', 2);

plt.xlabel('Deviation from '+' $N/2$'+ 'heads');
plt.ylabel('Probability density function') # Label
    probability density function

plt.xlim([-200, 200])

'''Finding the width'''
Max_value = max(histdata[0])
delta =Max_value/20

elements = np.where( np.abs(histdata[0]-Max_value/
    np.exp(1)) <delta)[0]

width_experimental.append(max(elements[1:]-elements
   [:-1])*BW) # This is the list of the widths found
    experimentally
width_analytical.append(2*N_samples/np.sqrt(2*
    N_samples+2))

plt.clf()
plt.plot(width_analytical,width_experimental,'r*')
plt.xlabel('Analytical width');
plt.ylabel('Experimental Width')
for xy in zip(width_analytical, width_experimental):

```

```

# <--
plt.annotate('(%s, %s)' % xy, xy=xy,
             textcoords='data') # <--

plt.savefig('widths.pdf')
plt.show()

```

The end result is the following plot

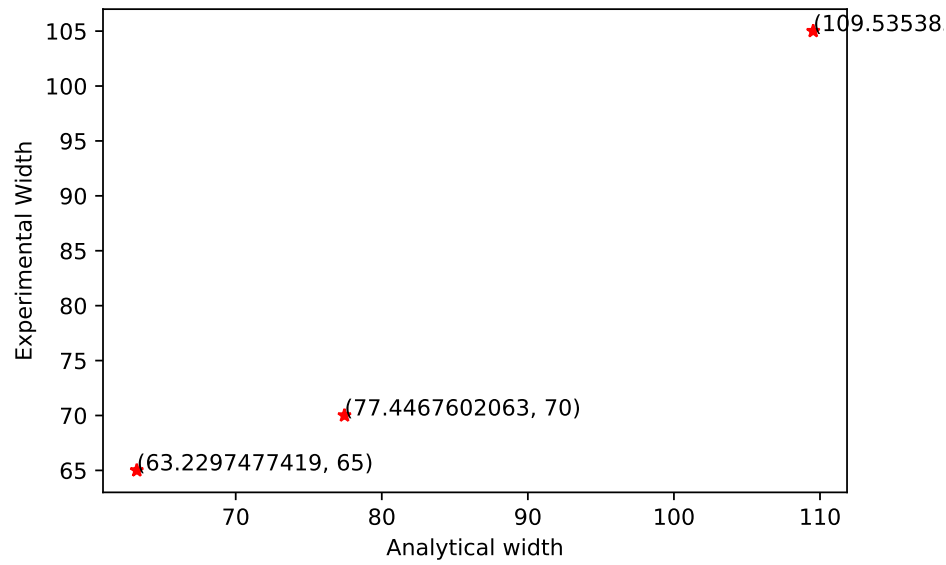


Figure 3: The widths for 3 distributions