1. There are $m=6n\ln(n)$ jobs assigned at random to n processors (note: remember $\ln(x)=\log_e(x)$). Consider a processor p and show, with high probability in parameter n, processor p does not receive more than $12\ln(n)$ jobs. (Hint: define an appropriate indicator variable for each job and apply the following Chernoff bound).

Theorem 1. Siano X_1, X_2, \ldots, X_n variabili indicatore indipendenti con $E[X_i] = p_i, 0 < p_i < 1$. Sia $X = \sum_{i=1}^n X_i$ $e \ \mu = E[X]$. Allora, per ogni $0 < \delta \le 1$,

$$Pr(X > (1+\delta)\mu) \le e^{-\frac{\delta^2 \mu}{3}}.$$

Here, we have X_i with $i=1,2...6n\ln(n)$, an indicator variable with value 1 if the i^{th} job gets assigned to processor p. Considering all events X_i are independent, so we get $\Pr(X_i=1)=\frac{1}{n}$.

Specifically, $X = \sum_{i=1}^{6n \ln(n)} X_i$. Given the processor has to receive no more than $12 \ln(n)$ jobs, this is our bound and in specific we have $12 \ln(n) = (1 + \delta)\mu$:

$$Pr(X > 12 \ln(n))$$

Now, we want to solve for δ :

$$12 \ln n = 6 \ln n (1 + \delta)$$
$$\frac{12 \ln n}{6 \ln n} = 1 + \delta$$
$$2 = 1 + \delta$$
$$\delta = 1$$

Then, we substitute everything:

$$\Pr(X > (1+1)\mu) = \Pr(X > 12\ln n) \le e^{-\frac{1^2 \cdot \mu}{3}}$$

$$\Pr(X > 12\ln n) \le e^{-\frac{6\ln n}{3}} = e^{-2\ln n} = \left(e^{\ln n}\right)^{-2} = (n)^{-2} = \frac{1}{n^2}$$

Thus we have:

$$\Pr(X > 12\ln n) \le \frac{1}{n^2}$$

which means:

$$\Pr(X \le 12 \ln n) \ge 1 - \frac{1}{n^2}$$

This shows that with high probability, the processor p does not receive more than $12\ln(n)$ jobs

2.

Exercise 2 (9 points) For $n \gg 1$, let X_1, X_2, \ldots, X_n be independent indicator random variables such that $Pr(X_i = 1) = (6 \ln n)/n$ (recall that $\ln n = \log_e n$). Let $X = \sum_{i=1}^n X_i$ and $\mu = E[X]$. By applying the following Chernoff bound

$$Pr(X > (1+\delta)\mu) < e^{-\mu\delta^2/2}$$
 for $0 < \delta \le 2e - 1$

prove that

$$Pr(X > 10 \ln n) < \frac{1}{n^c}$$

for some positive constant c to be determined.

Here, given $\Pr(X_i = 1) = \frac{6 \ln(n)}{n}$, we have:

$$\mu = E[X] = \sum_{i=1}^{n} E[X_i] = n \cdot \frac{6 \ln n}{n} = 6 \ln n$$

We then set the Chernoff bounds to be:

$$10\ln(n) = (1+\delta)6\ln(n)$$

Solve for δ :

$$\frac{10 \ln n}{6 \ln n} = 1 + \delta$$
$$\frac{10}{6} = 1 + \delta$$
$$\frac{5}{3} = 1 + \delta$$
$$\delta = \frac{5}{3} - 1 = \frac{2}{3}$$

Apply the Chernoff bound:

$$\Pr(X > 10 \ln n) = \Pr(X > (1 + \frac{2}{3})\mu)$$

Using the Chernoff bound formula:

$$\Pr(X > (1+\delta)\mu) < e^{-\frac{\mu\delta^2}{2}}$$

Substituting $\delta = \frac{2}{3}$ and $\mu = 6 \ln n$:

$$\Pr(X > (1 + \frac{2}{3})6 \ln n) < e^{-\frac{6 \ln n \cdot (\frac{2}{3})^2}{2}}$$

$$= e^{-\frac{6 \ln n \cdot \frac{4}{9}}{2}}$$

$$= e^{-\frac{6 \ln n \cdot \frac{4}{9}}{2}}$$

$$= e^{-\frac{24 \ln n}{18}}$$

$$= e^{-\frac{4 \ln n}{3}}$$

$$e^{-\frac{4\ln n}{3}} = (e^{\ln n})^{-\frac{4}{3}} = n^{-\frac{4}{3}}$$

Thus, we have:

$$\Pr(X > 10 \ln n) < \frac{1}{n^{4/3}}$$

Therefore, we have correctly proved that:

$$\Pr(X > 10 \ln(n)) < \frac{1}{n^{\frac{4}{3}}}$$

for $c = \frac{4}{3}$, bounding correctly for a positive constant c.

3.

Exercise 2 (11 points) Let S be a set of n distinct positive integers, and let WORK(S) be a procedure which, given input S, returns an integer by performing n^2 operations. Now consider the following randomized algorithm:

```
RAND_REC(S)
    if |S| <= 1 then return 1
    x = WORK(S)
    p = RANDOM(S)
     S1 = \{s \text{ in } S \text{ such that } s < p\}
     S2 = \{s \text{ in } S \text{ such that } s > p\}
     if (|S1| >= |S2|) then
         y = RAND_REC(S1)
     else
         y = RAND_REC(S2)
      return x + y
Applying the following Chernoff bound show that the complexity of RAND_REC(S) is O(n^2 \log n)
with high probability. (Hint: recall the analysis of randomized QuickSort.)
Theorem 1. Let X_1, X_2, \ldots, X_n be independent indicator random variables such that E[X_i] = p_i, 0 < \infty
p_i < 1. Let X = \sum_{i=1}^{n} X_i and \mu = E[X]. Then, for 0 < \delta \le 1,
                                      \Pr(X < (1 - \delta)\mu) < e^{-\mu \delta^2/2}.
```

The algorithm basically partitions the original sets into two subsets S_1 and S_2 containing elements from both sets, then calling the procedure on the larger set and returning the sum. We observe the following:

- if the sets are balanced S_1 and S_2 , then $O(\log(n))$
- we want to ensure that each recursive call reduces the size of the set by a significant fraction
 - o then consider a good pivot ensuring a probability of $\max(|S_1|, |S_2|) \le \frac{3n}{4}$

Using Chernoff bounds, we can show that the probability of having a "bad" pivot that doesn't split the set significantly decreases exponentially. For a random pivot p, the probability that the size of the larger partition is at most $\frac{3n}{4}$ can be derived as follows:

- Let X be the size of the larger partition.
- For $0 < \delta \le 1$, we have:

$$\Pr\left(X>(1+\delta)rac{n}{2}
ight)\leq e^{-rac{n}{2}\delta^2\over 3}$$

We want to ensure that the size of the larger partition is less than $\frac{3n}{4}$:

• Set
$$(1+\delta)\frac{n}{2}=\frac{3n}{4}$$

$$1+\delta=rac{3}{2} \implies \delta=rac{1}{2}$$

Using this value in the Chernoff bound:

•
$$\mu = \frac{n}{2}$$

$$\Pr\left(X>rac{3n}{4}
ight)\leq e^{-rac{n}{2}\left(rac{1}{2}
ight)^2}=e^{-rac{n}{24}}$$

This shows that the probability of a "bad" split is exponentially small.

The total work done at each level of recursion, WORK(S) performs n^2 operations.

Total work done is:

$$\sum_{i=0}^{O(\log(n)} n^2 = O(n^2 \log(n))$$

4.

Exercise 2 (9 points) Suppose you toss $n \gg 1$ times a coin: applying the following Chernoff bound show that the probability that you obtain more than $n/2 + \sqrt{6n \ln n}/2$ heads is at most 1/n.

Theorem 1. Let X_1, X_2, \ldots, X_n be independent indicator random variables such that $E[X_i] = p_i, 0 < p_i < 1$. Let $X = \sum_{i=1}^n X_i$ and $\mu = E[X]$. Then, for $0 < \delta \le 1$,

$$\Pr(X > (1+\delta)\mu) \le e^{-\mu\delta^2/3}.$$

Let X_i be an indicator random variable for the i-th coin toss, where $X_i=1$ if the outcome is heads and $X_i=0$ if the outcome is tails.

Since the coin is fair, $\Pr(X_i=1)=rac{1}{2}$. Therefore,

$$\mu = \mathbb{E}[X] = \sum_{i=1}^n \mathbb{E}[X_i] = n \cdot rac{1}{2} = rac{n}{2}$$

¹Recall that $\ln n = \log_e n$.

We need to find δ such that $rac{n}{2}+\sqrt{6n\ln\left(rac{n}{2}
ight)}=(1+\delta)rac{n}{2}$

1. Solving for δ :

$$egin{aligned} rac{n}{2} + \sqrt{6n\ln\left(rac{n}{2}
ight)} &= (1+\delta)rac{n}{2} \ & 1 + rac{2\sqrt{6n\ln\left(rac{n}{2}
ight)}}{n} &= 1 + \delta \ & \delta &= rac{2\sqrt{6n\ln\left(rac{n}{2}
ight)}}{n} \end{aligned}$$

Simplifying δ :

We substitute
$$\delta=2\sqrt{rac{6\ln\left(rac{n}{2}
ight)}{n}}$$
 and $\mu=rac{n}{2}$ into the Chernoff bound:

$$\Pr\left(X>rac{n}{2}+\sqrt{6n\ln\left(rac{n}{2}
ight)}
ight)=\Pr\left(X>(1+\delta)\mu
ight)\leq e^{-rac{\mu\delta^2}{3}}$$

Substituting μ and δ :

$$e^{-\frac{\frac{n}{2}\left(2\sqrt{\frac{6\ln\left(\frac{n}{2}\right)}{n}}\right)^2}}$$

$$=e^{-\frac{\frac{n}{2}\cdot\frac{24\ln\left(\frac{n}{2}\right)}{n}}{3}}$$

$$=e^{-\frac{12\ln\left(\frac{n}{2}\right)}{3}}$$

$$=e^{-4\ln\left(\frac{n}{2}\right)}$$

$$\delta = 2\sqrt{rac{6\ln\left(rac{n}{2}
ight)}{n}}$$

Since $\ln\left(\frac{n}{2}\right) = \ln n - \ln 2$:

$$e^{-4(\ln n - \ln 2)} = e^{-4\ln n + 4\ln 2}$$

Simplifying further:

$$= e^{-4 \ln n} \cdot e^{4 \ln 2}$$
$$= (e^{\ln n})^{-4} \cdot 2^4$$
$$= n^{-4} \cdot 16$$

Finally, we get:

$$\Pr\left(X > \frac{n}{2} + \sqrt{6n\ln\left(\frac{n}{2}\right)}\right) \leq \frac{16}{n^4}$$

For $n\gg 1$, $\frac{16}{n^4}<\frac{1}{n}$, so we have shown that:

$$\Pr\left(X > \frac{n}{2} + \sqrt{6n\ln\left(\frac{n}{2}\right)}\right) \leq \frac{1}{n}$$

Exercise 2 (10 points) Suppose you have a randomized algorithm for a minimization problem A that returns the correct output with probability at least 1/n, where n is the input size. Show how to obtain an algorithm for A that returns the correct output with high probability. (Hint: for the analysis use this inequality: $(1+x/y)^y \le e^x$ for $y \ge 1$, $y \ge x$.)

Characterize the event:

$$\Pr\left(X > \frac{1}{n}\right) > \frac{1}{n^d}$$

We want to find a value for k such that $Pr\left(1-\frac{1}{n}\right)^k \leq \frac{1}{n^d}$. In this case, it's standard the use of this inequality:

$$\left(1 + \frac{x}{y}\right)^y \le e^x, y \ge 1, y \ge x$$

Recall the following:

$$e^{-\ln(n^d)} = \frac{1}{n^d}$$

This inequality is derived from the exponential function and the binomial expansion. It represents an upper bound on the expression $\left(1+\frac{x}{y}\right)^y$, showing that it grows slower than e^x . The probability of not contracting the minimum cut in each iteration needs to be bounded and manipulated to ensure the overall algorithm's success probability is high.

Recall the following:

$$e^{-\ln(n^d)} = \frac{1}{n^d}$$

By choosing $k = \frac{dn}{2} \ln(n)$ it follows that:

$$\left(1 - \frac{1}{n^2}\right)^{k=n} \le e^{-1} = \frac{1}{e} \to \text{is } \underline{\text{not}} \text{ in the form } \frac{1}{n^d}$$

$$\left(\left(1 - \frac{1}{n}\right)^n\right)^{\ln(n^d)} = \left(1 - \frac{1}{m}\right)^{n\ln(n^d)}$$

Let's wrap up:

$$\left(1 - \frac{1}{n}\right)^{k = dn \ln(n^d)} = \left(\left(1 - \frac{1}{m}\right)^n\right)^{\ln(n^d)}$$

$$\leq (e^{-1})^{\ln(n^d)} = e^{-\ln(n^d)} = \frac{1}{n^d}$$

Problem 2 (10 points) Suppose you throw n balls into $\frac{n}{6 \ln n}$ bins¹ independently and uniformly at random. Applying the following Chernoff bound show that, with high probability, the bin with maximum load (load = number of balls in the bin) contains at most $12 \ln n$ balls. (Hint: focus first on one arbitrary bin and bound the probability of that bin's load exceeding $12 \ln n \ldots$)

Theorem 1. Let $X_1, X_2, ..., X_n$ be independent indicator random variables such that $E[X_i] = p_i, 0 < p_i < 1$. Let $X = \sum_{i=1}^n X_i$ and $\mu = E[X]$. Then, for $0 < \delta \le 1$,

$$\Pr(X > (1+\delta)\mu) \le e^{-\mu\delta^2/3}.$$

1. Define the Problem Setup:

- Number of balls, n.
- Number of bins, $m = \frac{n}{6 \ln n}$.
- · Each ball is independently and uniformly placed in one of the bins.

2. Random Variable Definition:

- Let X_i be the load (number of balls) in the i-th bin.
- X_i is a sum of independent indicator random variables, where each indicator variable denotes whether a particular ball lands in the i-th bin.

3. Expected Load Calculation:

- The probability that a particular ball lands in any specific bin is $\frac{1}{m}$.
- Thus, the expected number of balls in any bin (the expected load), μ , is:

$$\mu=rac{n}{m}=rac{n}{rac{n}{6\ln n}}=6\ln n$$

4. Applying the Chernoff Bound:

- We need to bound the probability that X_i exceeds $12 \ln n$.
- Let $X=X_{i}$, the load of a particular bin.
- ullet Using the Chernoff bound theorem, set δ such that:

$$(1+\delta)\mu = 12\ln n$$

• Since $\mu = 6 \ln n$:

$$(1+\delta)6\ln n = 12\ln n \implies 1+\delta = 2 \implies \delta = 1$$

$$\Pr(X > 12\ln n) = \Pr(X > (1+1)6\ln n)$$

$$\leq e^{-\frac{6\ln n}{3}}$$

$$= e^{-2\ln n}$$
$$= e^{\ln n^{-2}}$$

$$=1/n^2$$
.

We now show that, with high probability, the bin with maximum load contains at most $12 \ln n$ balls. Applying the union bound over all the $\frac{n}{6 \ln n}$ bins, we have that the probability that at least one bin gets more than $12 \ln n$ balls is at most

$$\frac{n}{6\ln n} \cdot \frac{1}{n^2} = \frac{1}{6n\ln n}.$$

In other words, the load of no bin exceeds $12 \ln n$ with probability at least $1 - 1/6n \ln n = 1 - o(1/n)$.

¹Recall that $\ln n = \log_e n$.