

$$1. P(\text{at least 2 have common birthday}) = 1 - P(\text{no common birthdays})$$

$$= 1 - (1) \left( \frac{364}{365} \right) \left( \frac{363}{365} \right) \dots \left( \frac{365 - (n-1)}{365} \right) = 1 - \frac{365!}{(365-n)! 365^n}$$

$$P(23) = 0.507 \quad (\text{See MATLAB code})$$

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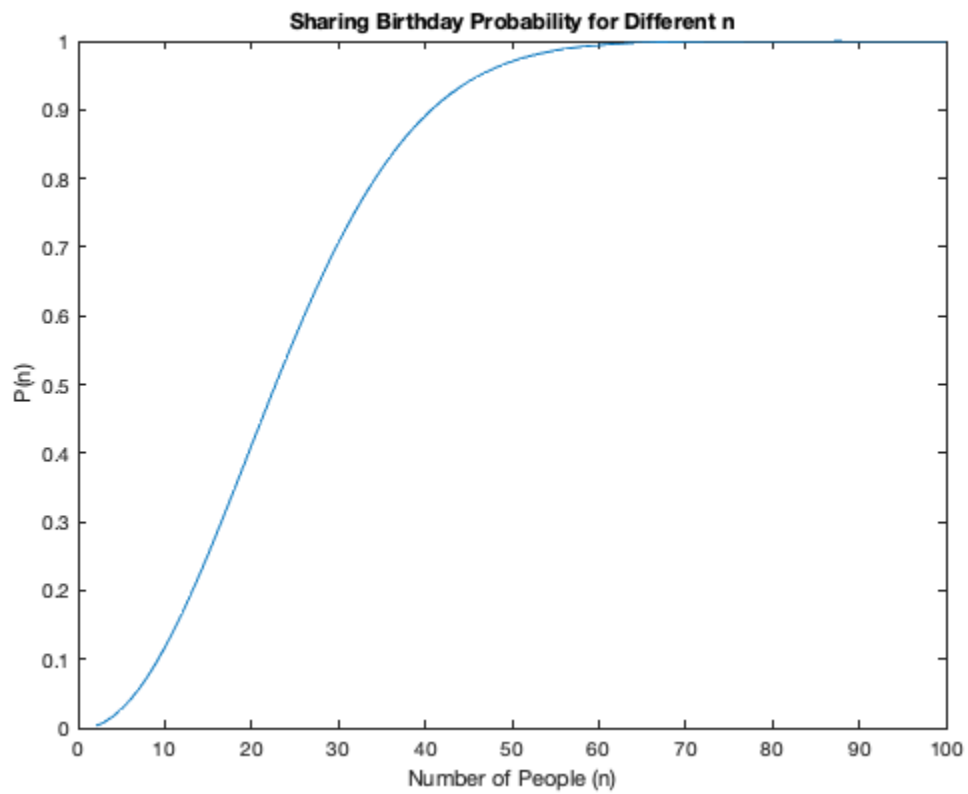
```
% populate x and y for plotting
x = 2:100;
y = [];
for i = 2:100
    y(i-1) = birthday_probability(i);
end

% plot
plot(x, y);
title("Sharing Birthday Probability for Different n");
xlabel("Number of People (n)");
ylabel("P(n)");

% print birthday probability for 23 people
fprintf("P(23): %d\n", birthday_probability(23));

% define function to return birthday probability for given n
function p = birthday_probability(n)
    product = 1.0;
    for i = 1:n
        product = product * (366 - i) / 365;
    end
    p = 1 - product;
end

P(23): 5.072972e-01
```



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$$\begin{aligned} 2. \quad P(\text{disease} \mid \text{tested positive}) &= \frac{P(\text{disease} \cap \text{tested positive})}{P(\text{tested positive})} \\ &= \frac{\frac{1}{1000} \cdot 0.99}{\frac{1}{1000} \cdot 0.99 + 999 \cdot \frac{1}{1000} \cdot 0.02} = 0.0472 \end{aligned}$$

3. a)  $E[X] = np$

b) see code.

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```

p = 0.3;
n = 100;
r = 10^3;

X = binornd(n, p, r, 1);

histogram(X, 20);
xline(30, '--r', {'E[X]'}, LineWidth=3);
title("Coin Tosses");
xlabel("Number of Heads");
ylabel("Occurrences");
legend("X", "E[X]")

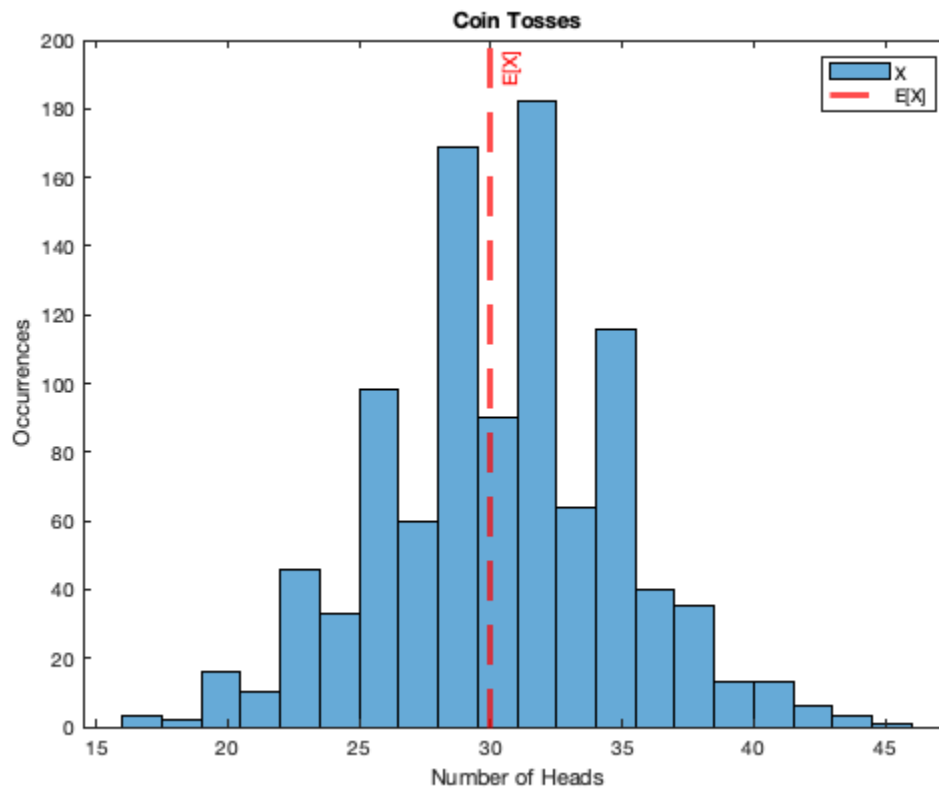
% compare average X to E[X]:
fprintf("Average number of heads: %.2f\n", mean(X));
fprintf("Expected number of heads: %d\n", n * p);

```

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Average number of heads: 30.13
Expected number of heads: 30

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4. See code.

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```
% define A, B, and AB
A = [2 4 6];
B = [1 2 3 4];
AB = [2 4];

% count occurrences
As = 0.0;
Bs = 0.0;
ABs = 0.0;

% run simulations
simulations = 10^4;
for i = 1:simulations
    die_roll = unidrnd(6);
    if (ismember(die_roll, A))
        As = As + 1;
    end
    if (ismember(die_roll, B))
        Bs = Bs + 1;
    end
    if (ismember(die_roll, AB))
        ABs = ABs + 1;
    end
end
As = As / simulations;
Bs = Bs / simulations;
ABs = ABs / simulations;

fprintf("P(A): %.3f\n", As);
fprintf("P(B): %.3f\n", Bs);
fprintf("P(AB): %.3f\n", ABs);
fprintf("P(A)P(B): %.3f\n", As*Bs);

P(A): 0.502
P(B): 0.663
P(AB): 0.337
P(A)P(B): 0.333
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$$3. a) Z \sim U[0,1], \quad X = F^{-1}(Z)$$

The CDF of  $X$  is

$$\begin{aligned} P(X \leq x) &= P(F^{-1}(Z) \leq x) \\ &= P(Z \leq F(x)) \\ &= F(x) \end{aligned}$$

since  $Z$  is a standard uniform distribution,  
 $P(Z \leq z) = z$  for any  $z \in [0,1]$ .

Hence,  $X \sim F$ .

$$b) X \sim \text{Exp}(\lambda)$$

$$F(x) = 1 - e^{-x/\lambda}$$

$$Z = 1 - e^{-x/\lambda}$$

$$e^{-x/\lambda} = 1 - Z$$

$$-\frac{x}{\lambda} = \ln(1 - Z)$$

$$x = -\lambda \ln(1 - Z)$$

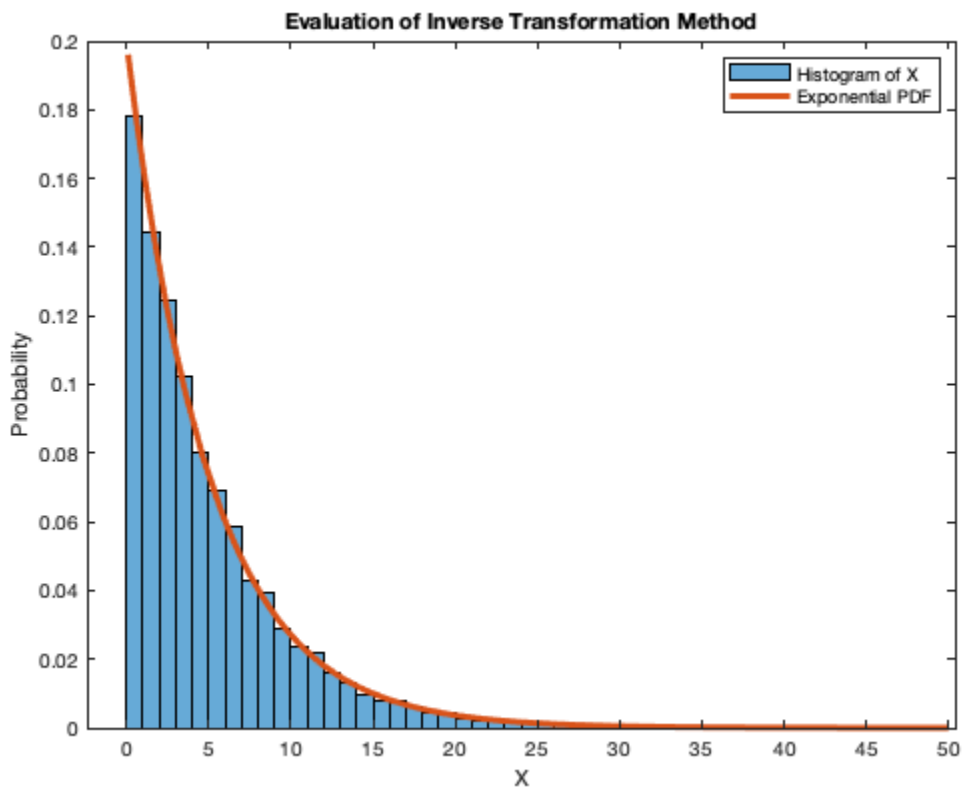
$$X = F^{-1}(Z) \Rightarrow F(x) = Z$$

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```
N = 10^4;
lambda = 5;

Z = rand(N, 1);
X = -lambda * log(1 - Z);

histogram(X, 'Normalization', 'pdf');
hold on
fplot(@(x) exppdf(x, lambda), [0.1, 50], LineWidth=3);
hold off
xlabel("X");
ylabel("Probability");
title("Evaluation of Inverse Transformation Method");
legend("Histogram of X", "Exponential PDF");
```



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6. Let  $L$  be the length of piece 1.

$$X = \max(L, 1-L)$$

$$X = \begin{cases} L & \text{if } L \geq \frac{1}{2} \\ 1-L & \text{if } L < \frac{1}{2} \end{cases}$$

$$E[X] = \int_0^{\frac{1}{2}} (1-l) dl + \int_{\frac{1}{2}}^1 l dl = 0.75$$

7. Algorithm I:

$$\begin{aligned} P(X=0) &= P(Y_2=0 \mid Y_1 \neq Y_2) = \frac{P(Y_2=0 \cap Y_1 \neq Y_2)}{P(Y_1 \neq Y_2)} \\ &= \frac{P(Y_2=0 \cap Y_1=1)}{P(Y_1 \neq Y_2)} = \frac{P(Y_2=0 \cap Y_1=1)}{P(Y_2=0 \cap Y_1=1) + P(Y_2=1 \cap Y_1=0)} \\ &= \frac{p(1-p)}{p(1-p) + (1-p)p} = \frac{1}{2} \end{aligned}$$

Since  $P(X=0) = \frac{1}{2}$  and  $P(X=1) = 1 - P(X=0) = \frac{1}{2}$ , Algorithm I does simulate a fair coin.

Algorithm II:

Let the outcome of the first toss be denoted  $Y_1$ .

If  $Y_1=1$ , then the coin will continue to be tossed until it lands on 0, and  $X$  becomes 0. Therefore, the tosses follow the form  $1 \dots 10$ , where there are  $k$  1's and  $k \geq 1$ . Mathematically,

$$P(X=0) = \sum_{k=1}^{\infty} (1-p)^k p$$

If  $Y_1=0$ , then the coin will continue to be tossed until it lands on 1, and  $X$  becomes 1. Therefore, the tosses follow the form  $00 \dots 01$ , where there are  $k$  0's and  $k \geq 1$ . Mathematically,

$$P(X=1) = \sum_{k=1}^{\infty} p^k (1-p)$$

Since  $p \neq 1-p$ ,  $\sum_{k=1}^{\infty} (1-p)^k p \neq \sum_{k=1}^{\infty} p^k (1-p)$ , so  $P(X=0) \neq P(X=1)$ .

Therefore, Algorithm II is not a simulation of an unbiased coin.

8. Let  $X$  = number of pokemon balls opened to have  $n$  different types  
Let  $X_i$  = number of pokemon balls opened to get the  $i$ th unique type  
s.t.  $X = X_1 + X_2 + X_3 + \dots + X_n$

$$P(\text{finding 1st unique pokemon at a given trial}) = \frac{n}{n}$$

$$P(\text{finding 2nd unique pokemon at a given trial}) = \frac{n-1}{n}$$

:

$$P(\text{finding } i^{\text{th}} \text{ unique pokemon at a given trial}) = \frac{n+1-i}{n}$$

$$X_i \sim \text{Geo}(P(\text{finding } i^{\text{th}} \text{ unique pokemon at a given trial})) = \text{Geo}\left(\frac{n+1-i}{n}\right)$$

$$E[X_i] = \frac{n}{n+1-i}$$

$$E[X] = E[X_1 + X_2 + \dots + X_n] = E[X_1] + E[X_2] + \dots + E[X_n]$$

$$= \sum_{i=1}^n \frac{n}{n+1-i}$$

$$= \sum_{i=0}^{n-1} \frac{n}{n-i}$$