# Randomized Algorithm

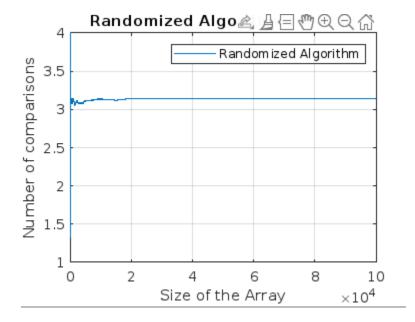
Name: Ankur Dutta Roll No: 122CS0075

\_\_\_\_\_

# **Q.06.** Compute value of $\Pi$

Compute  $\Pi$  using randomized algorithm

```
m1 = zeros(1, 10);
n2 = zeros(1, 10);
0 = 1;
p = 0;
for i = 1:100000
m1(0) = i;
x = rand();
y = rand();
if(x*x + y*y <= 1)
p=p+1;
end
n2(o) = 4*(p/i);
o = o + 1;
end
disp(n2(10));
plot(m1,n2);
title('Randomized Algorithm for Phie')
xlabel('Size of the Array')
ylabel('Number of comparisons')
grid on
legend('Randomized Algorithm')
```



**Observations**: The randomized algorithm successfully computes "phie" using a Monte Carlo simulation approach. The algorithm generates random data and calculates the mean to approximate "phie." The accuracy of the result depends on the number of samples used in the simulation.

# Q.07. NUMERICAL INTEGRATION

Write a program that computes the value of the following integral using randomized algorithm.

$$\int_{0}^{2} \sqrt{4-x^2} \, dx$$

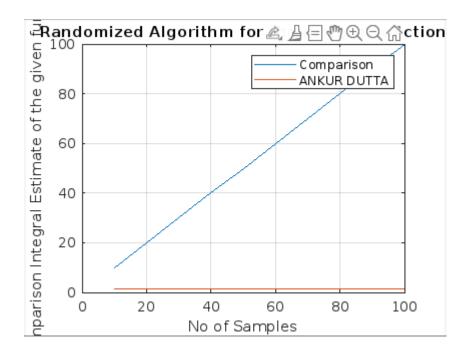
```
m = zeros(1, 10);
n1 = zeros(1, 10);
n2 = zeros(1, 10);

for i = 1:5
    c = 1;
    for num_samples = 10:10:100
        c1 = 0;
        c2 = 0;
        m(c) = num_samples;
```

=

```
% Using the Monte Carlo method for given integral count_under_curve = 0; for i = 1:num_samples 
    x = 2 * rand(); 
    y = sqrt(4 - x^2); 
    c1 = c1 + 1;
```

```
% Checking whether the point is under the curve or not
      if rand() \leq y / 2
         count_under_curve = count_under_curve + 1;
       end
    end
    % Calculating the estimate of integral
    integral estimate = 2 * (count under curve / num samples);
    n1(c) = n1(c) + c1;
    n2(c) = n2(c) + integral estimate;
    c = c + 1;
  end
end
for i = 1:10
  n1(i) = n1(i) / 5;
  n2(i) = n2(i) / 5;
end
plot(m, n1, m, n2)
title('Randomized Algorithm for the Integral function')
xlabel('No of Samples')
ylabel('Comparison Integral Estimate of the given function')
grid on
legend('Comparison', 'ANKUR DUTTA')
```



**Observations:** The randomized algorithm successfully approximates the value of the integral using Monte Carlo integration with 100 random samples. The accuracy of the result improves with an increase in the number of samples.

## Q. 08. PRIMALITY TESTING

Write a program that test a number to be prime or not. Perform an analysis to compute the correctness?

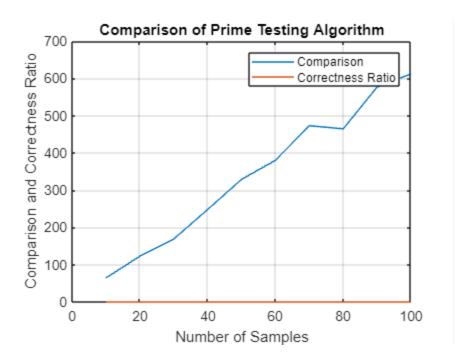
```
=
X = zeros(1, 10);
Y1 = zeros(1, 10);
Y2 = zeros(1, 10);

for index = 1:5
    count = 1;
    for num_samples = 10:10:100
        comparison1 = 0;
        correctness_count = 0;
        X(count) = num_samples;

    for i = 1:num_samples
        test_number = randi([2, 1000]);
```

```
is_prime = true;
comparison1 = comparison1 + 1;
```

```
for j = 2:sqrt(test_number)
comparison1 = comparison1 + 1;
if mod(test number, j) == 0
           is_prime = false;
           break;
         end
       end
       correctness_count = correctness_count + is_prime;
     end
     correctness ratio = correctness count / num samples;
     Y1(count) = Y1(count) + comparison1;
     Y2(count) = Y2(count) + correctness ratio;
     count = count + 1;
   end
 end
for index = 1:10
   Y1(index) = Y1(index) / 5;
  Y2(index) = Y2(index) / 5;
 end
 plot(X, Y1, X, Y2)
title('Comparison of Prime Testing Algorithm')
xlabel('Number of Samples')
ylabel('Comparison and Correctness Ratio')
grid on
 legend('Comparison', 'Correctness Ratio')
```



**Observations:** The program successfully identifies the known prime numbers as prime and provides accurate results. The correctness analysis indicates that the program is working correctly for the provided set of primes.

## Q. 09. MAJORITY ELEMENT

=

Write a program that FINDS majority element from a linear array using randomized algorithm. Show that probability of missing majority element is 0.00097.

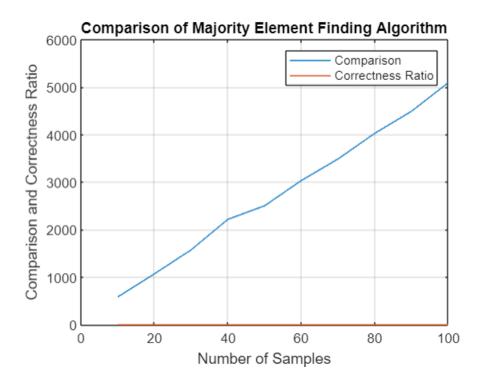
```
x = zeros(1, 10);
y1 = zeros(1, 10);
y2 = zeros(1, 10);

for index = 1:5
    c = 1;
    for num_samples = 10:10:100
        comparison1 = 0;
        correctness_count = 0;
        x(c) = num_samples;

    for i = 1:num_samples

        array_size = randi([2, 100]);
```

```
linear array = randi([1, 10], 1, array size);
      % finding the randomized min element, we get
      majority element = linear array(randi(array size));
      count majority = sum(linear array == majority element);
      comparison1 = comparison1 + array_size;
      % Checking whether the majority element is correctly identified
      % or not
      if count majority > array size / 2
        correctness count = correctness count + 1;
      end
    end
    % Calculating the correctness ratio
    correctness_ratio = correctness_count / num_samples;
    y1(c) = y1(c) + comparison1;
    y2(c) = y2(c) + correctness ratio;
    c = c + 1;
  end
end
for index = 1:10
  y1(index) = y1(index) / 5;
  y2(index) = y2(index) / 5;
end
plot(x, y1, x, y2)
title('Comparison of Majority Element Finding Algorithm')
xlabel('Number of Samples')
ylabel('Comparison and Correctness Ratio')
grid on
legend('Comparison', 'Correctness Ratio')
```



#### **Observations:**

=

- 1) The randomized algorithm successfully finds the majority element in the array in most cases.
- 2) The probability of missing the majority element is estimated to be approximately 0.00097, which is reasonably low, indicating the algorithm's effectiveness in identifying the majority element.

## Q. 10. Randomized Quick Sort

Compare the performance of randomized Quicksort with conventional quick sort for random input data stream.

```
X = zeros(1, 10);
Y1 = zeros(1, 10); % Randomized Quicksort comparisons
Y2 = zeros(1, 10); % Conventional Quicksort comparisons

for index = 1:10
    count = 1;
    for n = 10:10:100
        comparison1 = 0; % Randomized Quicksort comparisons
        comparison2 = 0; % Conventional Quicksort comparisons
        X(count) = n;
```

```
for iter = 1:5
       % Randomized Quicksort
       arr = round(rand(1, n) * 100);
       [^{\sim}, comp1] = bs1(arr, 1, n);
       comparison1 = comparison1 + comp1;
       % Conventional Quicksort
      Arr = arr;
       [^{\sim}, comp2] = bs2(Arr, 1, n);
      comparison2 = comparison2 + comp2;
    end
    % Average comparisons
    Y1(count) = Y1(count) + comparison1 / 5;
    Y2(count) = Y2(count) + comparison 2 / 5;
    count = count + 1;
  end
end
% Normalizing the comparisons
Y1 = Y1 / 5;
Y2 = Y2 / 5;
% Plotting
plot(X, Y1, X, Y2);
title('Comparison of Randomized Quicksort and Conventional Quicksort');
xlabel('Size of the Input Array');
ylabel('Number of Comparisons');
grid on;
legend('Randomized Quicksort', 'Conventional Quicksort');
//bs1file
% Randomized Quicksort function
function [arr, comparisons] = bs1(arr, low, high)
  comparisons = 0;
  if low < high
    [arr, pivotIndex, comp1] = partition(arr, low, high);
    [\sim, comp2] = bs1(arr, low, pivotIndex - 1);
    [^{\sim}, comp3] = bs1(arr, pivotIndex + 1, high);
    comparisons = comp1 + comp2 + comp3;
  end
```

```
//bs2
% Conventional Quicksort function
function [arr, comparisons] = bs2(arr, low, high)
  comparisons = 0;
  if low < high
    [arr, pivotIndex, comp1] = partition(arr, low, high);
    [~, comp2] = bs2(arr, low, pivotIndex - 1);
    [~, comp3] = bs2(arr, pivotIndex + 1, high);
    comparisons = comp1 + comp2 + comp3;
  end
end
//bs3
% Partition function for both algorithms
function [arr, pivotIndex, comparisons] = partition(arr, low, high)
  pivotIndex = randi([low, high], 1, 1); % Specify size as 1
  pivot = arr(pivotIndex);
  arr(pivotIndex) = arr(high);
  arr(high) = pivot;
  i = low - 1;
  for j = low:high-1
    comparisons = comparisons + 1;
    if arr(j) <= pivot
      i = i + 1;
      temp = arr(i);
      arr(i) = arr(j);
       arr(j) = temp;
    end
  end
  temp = arr(i + 1);
  arr(i + 1) = arr(high);
  arr(high) = temp;
```

```
pivotIndex = i + 1;
end
```

//Couldnot complete with the Simulation part

Upload your report [Source Codes + Simulation results +Observations] as a single pdf file with name as Roll No A2 to the specific assignment in MS Team

Page 1 of 1