

## Homework #4

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### Question 1

Write an algorithm that uses the dart-throwing method to approximate pi using a quasi-Monte Carlo method

```
% Number of Points
npoints = 10000000;

% Draw npoints Neiderrieter points on the unit square:
[points, weights] = qnwequi(npoints, [0 0], [1 1], ...
    'N');

% Evaluate the function at these points

fun __ evals = 4*ind __ function(points);

% Average the function evaluations

pi __ qmc = mean(fun __ evals)
```

```
pi __ qmc = 3.1416
```

### Question 2

```
% Write an algorithm that uses the dart throwing ...
% method to approximate pi
% using a Newton-Cotes method

% We are going to use the trapezoid rule

% Number of nodes in the x direction
xnodes = 10000;

% Number of nodes in the y direction
ynodes = 10000;

% Obtain nodes and weights
[points, weights] = qnwtrap([xnodes ynodes], [0 0], ...
    [1 1]);
```

```
% Approximate pi
pi_nc = weights '*4*ind_function(points)
```

```
pi_nc = 3.1416
```

### Question 3

```
% Approximate pi based on the above integral using ...
% a quasi-Monte Carlo
% approach
```

```
% Draw equidistributed sequence
```

```
% Number of Points
npoints = 10000000;
```

```
% Draw npoints Neiderrieter points on the unit ...
% interval:
[points, weights] = qnwequi(npoints, 0, 1, 'N');
```

```
% Evaluate the function at these points
```

```
fun_evals = integrand(points);
```

```
% Average the function evaluations
```

```
pi_qmc = mean(fun_evals)
```

```
pi_qmc = 3.1416
```

### Question 4

Write an algorithm that uses the dart-throwing method to approximate pi using a Newton-Cotes approach to computing integrals

```
% We are going to use the trapezoid rule
```

```
% Number of nodes in the x direction
xnodes = 10000000;
```

```
% Obtain nodes and weights
[points, weights] = qnwtrap(xnodes, 0, 1);
```

```
% Approximate pi
pi_nc = weights'*integrand(points)
```

```
pi_nc = 3.1416
```

### Question 5

Prepare a table which shows the mean-squared error of 200 simulations of pseudo-MC integration using 100, 1000, and 10,000 draws. Compare this to the squared error of the quasi-MC and Newton-Coates methods for the same number of quadrature draws (i.e. nodes).

#### For 100 draws

```
ndraws = 100;

% Number of Simulations
ns = 200;

% Vector for Storing integral estimates of each ...
simulation
pi_mc_100 = zeros(ns,1);

for i = 1:ns
    % Draw 100 pseudo-random numbers
    [nodes, weights] = qnwequi(ndraws, [0 0], [1 1], ...
        'R');
    % Evaluate the pi approximation
    pi_mc_100(i) = weights'*4*ind_function(nodes);
end

% Mean Squared Error
mse_100 = mean ( ( pi_mc_100 - pi ).^2 )
```

```
mse_100 = 0.0277
```

#### For 1000 draws

```
ndraws = 1000;

% Number of Simulations
ns = 200;

% Vector for Storing integral estimates
```

```

pi_mc_1000 = zeros(ns,1);

for i = 1:ns
    % Draw 1000 pseud0-random numbers
    [nodes, weights] = qnwequi(ndraws, [0 0], [1 1], ...
        'R');
    % Evaluate the indicator function at these nodes
    pi_mc_1000(i) = weights'*4*ind_function(nodes);
end

% Mean Squared Error
mse_1000 = mean ( ( pi_mc_1000 - pi ).^2 )

mse_1000 = 0.0025

```

### For 10,000 draws

```

ndraws = 10000;

% Number of Simulations
ns = 200;

% Vector for Storing integral estimates
pi_mc_10000 = zeros(ns,1);

for i = 1:ns
    % Draw 100 pseudo-random numbers
    [nodes, weights] = qnwequi(ndraws, [0 0], [1 1], ...
        'R');
    % Evaluate the indicator function at these nodes
    pi_mc_10000(i) = weights'*4*ind_function(nodes);
end

% Mean Squared Error
mse_10000 = mean ( ( pi_mc_10000 - pi ).^2 )

mse_10000 = 2.8866e-04

```

### Calculate squared errors for quasi-Monte Carlo and Newton-Coates

```

% Vector for Storing Squared Errors of quasi-Monte ...
% Carlo
sqe_qmc = zeros(3,1);

```

```

% First do quasi-Monte Carlo with 100 nodes
nodes = 100;

% Starting Position for storing Squared Errors
i = 1;

while nodes <= 10000

    % Draw npoints Neiderrieter points on the unit ...
    square:
    [points, weights] = qnwequi(nodes, [0 0], [1 1], ...
        'N');

    % Evaluate the function at these points
    fun_evals = 4*ind_function(points);

    % Obtain pi approximation
    pi_qmc = mean( fun_evals );

    % Calculate the squared error of the current ...
    integration approximation
    sqe_qmc(i) = (pi_qmc - pi)^2;

    % Increase the number of nodes by an order of ...
    magnitude
    nodes = 10*nodes;

    % Increase counter variable i
    i = i + 1;

end

```

## Squared Errors for Newton-Coates

```

% Vector for Storing Squared Errors of Newton-Coates
sqe_nc = zeros(3,1);

% First do Newton-Coates with 100
nodes = 100;

% Starting position for storing squared errors
i = 1;

```

```

while nodes <= 10000

    % Number of nodes in the x direction
    xnodes = nodes;
    % Number of nodes in the y direction
    ynodes = nodes;

    % Obtain nodes and weights
    [points, weights] = qnwtrap([xnodes ynodes], [0 ...
        0], [1 1]);

    % Approximate pi
    pi_nc = weights'*4*ind_function(points);

    % Calculate the squared error of the current ...
    % integration approximation
    sqe_nc(i) = ( pi_nc - pi ) ^ 2;

    % Increase the number of nodes by an order of ...
    % magnitude
    nodes = 10*nodes;

    % Increase counter variable
    i = i + 1;

end

```

## Display Tables comparing Mean Squared Errors

```

mse_mc = [mse_100; mse_1000; mse_10000];

varNames = {'MonteCarlo ', 'QuasiMonteCarlo ', ...
    'NewtonCoates '};

rowNames = {'100 ', '1000 ', '10000 '};
disp('Mean Squared Error Comparison');

```

Mean Squared Error Comparison

```

MSE = table(mse_mc, sqe_qmc, sqe_nc, ...
    'VariableNames', varNames, 'RowNames', rowNames)

```

	MonteCarlo	QuasiMonteCarlo	NewtonCoates
1 100	0.0277	4.6624e-04	1.3305e-05
2 1000	0.0025	2.5365e-06	2.4940e-08
3 10000	0.0003	6.2830e-07	1.1405e-11