

Due: *Friday, December 4 at 5 PM.*

Upload your solution to Stellar as a zip file “YOURNAME\_ASSIGNMENT\_9.zip” which includes the script for each question in the proper format *as well as* any MATLAB functions (of your own creation) your scripts may call. Remember to verify your upload before the assignment due time by downloading your solution folder yourself and re-running `grade_o_matic`. Full instructions for assignment preparation are given in the document “Preparing assignments for `grade_o_matic`” available on the 2.086 website.

In this assignment, the student mode of `grade_o_matic` will only provide partial credit for questions 1 and 5. No partial credit will be provided for the rest of the questions. This means that `grade_o_matic` will **not** check the MATLAB syntax of your scripts for questions 2 and 6, hence you are responsible for ensuring that your scripts have no syntax errors and that they perform correctly for any set of input data. *In fact, for all questions you should yourself devise several test cases for which you can anticipate the correct answers and hence test your script.*

For question 3, please submit your answer in pdf format. Your written responses may be typed in your favorite software, but scans of handwritten material will not be accepted.

## Questions

- (25 points) We would like you to write a function which computes a Monte Carlo estimate  $(\hat{A}_D)_n$ , and associated 95% confidence-level (normal-approximation two-sided) confidence interval  $[ci_{A_D}]_n$ , for the area of the region

$$D = \{x_1^2 + x_2^2 \leq 0.75\} \cup \{(x_1 - \alpha)^2 + x_2^2 \leq 0.75\}. \quad (1)$$

Note that  $D$  is the union of two circles in which the center of the second circle is shifted by  $\alpha$  (in  $x_1$ ) from the center of the first circle. We choose for our background rectangle  $R$  the square  $R = [-c, c]^2$  for  $c$  sufficiently large to enclose  $D$ .

Your function should have signature

```
function [area_est,area_conf_int] = MC_area(alpha,c,n,x1pts,x2pts)
```

with five function inputs and two function outputs. The first function input is a scalar `alpha`, which corresponds to the center shift of the second circle  $\alpha$ ; the allowable instances, or parameter domain, is  $0 \leq \alpha \leq 0.1$ . The second function input is a scalar `c` which corresponds to  $c$  that defines the bounding square  $R$ : the lower left corner of  $R$  is  $(x_1, x_2) = (-c, -c)$  and the upper right corner of  $R$  is  $(x_1, x_2) = (c, c)$ . The allowable instances are  $1 \leq c \leq 10$ . (Note that for these instances of `c` the domain  $D$  is indeed included in  $R$  for all allowable instances of `alpha`.) The third function input is a scalar integer `n`, which corresponds to the number of random darts; the allowable instances, or parameter domain, is  $n \in \{10, 11, \dots, 10^6\}$ . The fourth and fifth function inputs are the  $n \times 1$  vectors `x1pts` and `x2pts`, which correspond to the coordinates of the  $n$  random darts,  $(x1pts(i), x2pts(i))$ ,  $i \in \{1, 2, \dots, n\}$ , thrown at the square

$R$  — on the basis of which you will calculate your area estimate and associated confidence interval: we emphasize that `x1pts` and `x2pts` correspond to arrays of i.i.d. pseudo-random variates from the (univariate) continuous uniform distribution over the interval  $(-c, c)$ .

The first function output is a scalar `area_est` which corresponds to the Monte-Carlo estimate for the area,  $(\hat{A}_D)_n$ . The second function output is the  $1 \times 2$  row vector `area_conf_int` which corresponds to the confidence interval  $[ci_{A_D}]_n$  such that the entries `area_conf_int(1)` and `area_conf_int(2)` are respectively the lower and upper limits of the confidence interval  $[ci_{A_D}]_n$ .

Three further points: first, in the event that the normal-approximation criterion<sup>1</sup>,  $n\theta > \phi$  **and**  $n(1 - \theta) > \phi$  with  $\phi = 10$ , for the construction of the confidence interval is not satisfied *your code should return [-1,1]* for `area_conf_int`. Second, you should set  $z_\gamma = 1.96$  which corresponds to confidence level  $\gamma = 95\%$ . Third, in order to test your function `MC_area` for any desired `alpha`, `c` and `n`, you must first generate the random darts `x1pts` and `x2pts` *outside* your function. Note, however, that for purposes of grading, `grade_o_matic` will automatically generate appropriate inputs `x1pts`, `x2pts` for your function — no action needed on your part.

The script template for this question is provided in `A9Q1_Template`; no modifications are required (except to remove the `_Template` from the name). The function template is provided in `MC_area_Template.m`; note for function templates (as well as script templates) you should first save as with `_Template` removed from the name.

2. (20 points) We would like you to write a function which computes, based on the sample-mean method, a Monte Carlo estimate  $\hat{I}_n$  for the integral

$$I = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |\vec{x} - \vec{y}| \exp\left(-\frac{x^2}{2}\right) \exp\left(-\frac{y^2}{2}\right) d\vec{x} d\vec{y} \quad (2)$$

Here  $\vec{x}$  denotes the vector  $(x_1, x_2, x_3)$  with magnitude  $x = \sqrt{x_1^2 + x_2^2 + x_3^2}$ ; similar definitions hold for  $\vec{y}$ . It follows that  $d\vec{x}$  is the differential element  $dx_1 dx_2 dx_3$ ; similarly for  $d\vec{y}$ . As a result, the domain of integration is six-dimensional.

The domain of integration in equation (2) is infinite. To avoid the difficulty associated with discretizing an infinite domain, we will truncate it to the domain  $-5 \leq x_1, x_2, x_3, y_1, y_2, y_3 \leq 5$ , since the exponential function in the integrand makes the contribution of points beyond this truncated domain negligible. In other words, we will approximate (2) by

$$I \approx \int_{-5}^5 \int_{-5}^5 \int_{-5}^5 \int_{-5}^5 \int_{-5}^5 \int_{-5}^5 |\vec{x} - \vec{y}| \exp\left(-\frac{x^2}{2}\right) \exp\left(-\frac{y^2}{2}\right) dx_1 dx_2 dx_3 dy_1 dy_2 dy_3 \quad (3)$$

Write a function with signature

```
function [int_est] = sample_mean_integral(n,upts)
```

which takes two function inputs and yields one function output. The first function input is the scalar integer `n`, which is related to the size of the random sample; the allowable instances,

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<sup>1</sup>In general, the larger the value of  $\phi$  the better the approximation becomes;  $\phi \approx 5$  ensures errors of a few percent, while the value chosen here,  $\phi = 10$ , ensures errors on the order of 1% or less

or parameter domain, is  $\mathbf{n} \in \{10, 11, \dots, 10^5\}$ . The second function input is the  $\mathbf{n} \times 6$  array `upts`, which corresponds to i.i.d. pseudo-random variates from the (univariate) continuous uniform distribution over the interval  $(0, 1)$ . Every column of `upts` contains  $\mathbf{n}$  pseudo-random variates; columns 1 – 6 in `upts` will be related to  $x_1, x_2, x_3, y_1, y_2$  and  $y_3$ , respectively, after the transformations required for the integral estimation. The function yields a single function output: the scalar `int_est`, which is the sample-mean estimate for  $I, \hat{I}_n$ , based on (3).

One further point: in order to test your function `sample_mean_integral` you must first generate the random sample `upts` **outside** your function. Note, however, that for purposes of grading, `grade_o_matic` will automatically generate appropriate inputs `upts` for your function — no action needed on your part. **Attention:** `upts` is drawn from the continuous uniform distribution over the interval  $(0, 1)$ ; inside your function you must transform these pseudo-random variates to the correct interval as required in the integral (3).

The script template for this question is provided in `A9Q2_Template`; no modifications are required (except to remove the `_Template` from the name). The function template is provided in `sample_mean_integral_Template.m`.

3. (15 points) As you very well know, your estimate of the integral in question 2, `int_est`, is itself a random variable since it is a sum of random variables. Using your function from question 2 generate  $m = 5000$  realizations of this random variable and study its distribution for  $n = 10000$  and  $n = 50000$ . If your computer has multiple cores you may want to take advantage of the command `parfor` to reduce the computational cost of this exercise.

Your solution will not be graded by `grade_o_matic`; as a result, there will be no student instances or partial credit. Please submit your answer in a pdf document named `A9Q3.pdf`; this file should include two histograms of the distribution of the variable `int_est` (one for each of the cases  $n = 10000$  and  $n = 50000$ ), along with a discussion of your study of these two distributions (e.g. What is the observed distribution of `int_est`? Is it what you expected from theory? How does the standard deviation of `int_est` scale with  $n$ ? Is this what you expected?).

4. (15 points) Consider the following MATLAB script

```
% begin script

clear
% note we "clear" the workspace

n = 10000; % number of random darts

u1 = rand(1,n);
u2 = rand(1,n);

x1 = u1;
x2 = 2.*u2;

addfac = 1.0;

% BEGIN BLOCK
```

```

numinside = 0;
for i = 1:n
    if( x2(i) <= my_func(x1(i),addfac) )
        numinside = numinside + 1;
    end
end
% END BLOCK

```

```

area_estimate = (numinside/n)*2.0;

```

```

% end script

```

and function my\_func

```

function [value] = my_func(x,yplus)
value = x.^3 + yplus;
return
end

```

for approximating an area  $A_D$  of a domain  $D$  by the Monte Carlo method. In the limit that  $n$  tends to infinity, `area_estimate` will approach  $A_D$ .

- (i) (5 points) Each of the bivariate uniform random variates (realizations of random variables from the bivariate uniform density)  $(x1(i), x2(i))$ ,  $i = 1, \dots, n$ , will reside in the rectangle:

- (a)  $-1 < x1(i) < 1, -1 < x2(i) < 1$
- (b)  $-1 < x1(i) < 1, 0 < x2(i) < 1$
- (c)  $0 < x1(i) < 1, -1 < x2(i) < 1$
- (d)  $0 < x1(i) < 1, 0 < x2(i) < 2$

- (ii) (5 points) The area  $A_D$  is given by:

- (a)  $2/3$
- (b)  $4/3$
- (c)  $2$
- (d)  $5/4$

*Hint:* draw a sketch in which you indicate the rectangle over which  $x1, x2$  is defined (i.e., at which you throw your darts); then include in your sketch the domain  $D$  defined by the `if` condition in BLOCK; finally, recall the area interpretation of the definite integral to determine  $A_D$ .

- (iii) (5 points) Which single line of MATLAB code below is equivalent to (and also much more efficient than) the entire BLOCK of code in the script above:

- (a) `numinside = sum( find( x2 <= my_func(x1,addfac) ) );`
- (b) `numinside = sum( x2 <= my_func(x1,addfac) );`

```
(c) numinside = length( x2 <= my_func(x1,addfac) );
(d) numinside = length( my_func(x1,addfac) );
```

Note by the entire BLOCK of code we refer to the lines of code between the comments BEGIN BLOCK and END BLOCK.

We provide you with the template `A9Q4_Template.m` file that contains the multiple-choice format required by `grade_o_matic`.

5. (10 points) In Assignment 8, question 4, we asked you to create a function `positive_normal` which generates pseudo-random variates from a rectified normal distribution with mean `mu` and standard deviation `sigma`.

In this question, we ask you to modify `positive_normal` from Assignment 8 so that it generates random variates from a "rectified and bounded normal" distribution for which the probability of observing values below 0 and above 100 is zero. In other words, create a new function with signature:

```
function [xpts_r_and_b] = r_and_b_normal(mu,sigma,n)
```

which provides a row vector of `n` rectified and bounded normal random variates (which take values between 0 and 100, as explained above) originating from a normal population with mean `mu` and standard deviation `sigma` (note that the mean and standard deviation are defined before the bounding process).

The function takes three inputs. The first two inputs, respectively `mu` and `sigma`, are scalars; the allowable instances, or parameter domain, is  $0 \leq \mu \leq 100$  and  $1 \leq \sigma \leq 30$ . The input `n` is a scalar integer; the allowable instances, or parameter domain, is  $n \in \{10, 11, \dots, 10^5\}$ . The function yields a single function output: the output is the  $1 \times n$  row vector `xpts_r_and_b`, which is our pseudo-random sample realization. Note that `xpts_r_and_b` must be random not just in terms of the outcome frequencies but also in terms of the order - such that (say) the first `n/2` elements of `xpts_r_and_b`, or the last `n/2` elements of `xpts_r_and_b`, or the "middle" `n/2` elements of `xpts_r_and_b`, should also yield (approximately) the correct outcome frequencies.

The script for this question is provided in `A9Q5_Template`; no modifications are required (except to remove the `_Template` from the name). The function template is provided in `r_and_b_normal_Template.m`.

6. (15 points)

A 2.086 student hears in lab that Monte Carlo methods can be used for a lot more than just evaluating integrals. Having been preoccupied with their grade in 2.086, they decide to use Monte Carlo *simulation* to estimate the probability of getting an "A" in class. Their plan is as follows: they know that they have received already  $X$  credit based on 7 assignments and the midterm. Their final score will thus be

$$S = X + w_{A8}A8 + w_{A9}A9 + w_{IE}IE + w_{FE}FE \quad (4)$$

where  $A8$ ,  $A9$ ,  $IE$  and  $FE$  refer to the (yet unknown) scores in Assignment 8, Assignment 9, Instructor Evaluation and the Final Exam, and  $w_{A8}$ ,  $w_{A9}$ ,  $w_{IE}$  and  $w_{FE}$  to the corresponding weights, respectively. Note that here scores, including  $X$ , will be reported on the scale  $[0, 100]$ .

Given reliable stochastic data for these last 4 variables, one can generate stochastic data for their score in the class,  $S$ , which can be used to estimate the probability of getting any grade. In order to see how well this idea works, we ask you to create a function with signature

```
function [probs] = grade_distribution(r_and_b_grades,cutoffs,w,X)
```

which will return the  $5 \times 1$  vector **probs** containing the probabilities of the student obtaining the grades A, B, C, D and F (in that order) expressed as a fraction of unity (i.e. 0.95 for a probability of 95%). The inputs to this function are:

- The  $4 \times n$  array of bounded and rectified normal random variates **r\_and\_b\_grades**. Each row of this array contains **n** random variates corresponding to **n** realizations of the student performance in Assignment 8, Assignment 9, Instructor Evaluation and the Final Exam, respectively.
- The  $4 \times 1$  vector **cutoffs** containing the *lower* cutoff for each of the grades A, B, C and D (in that order) expressed on the scale  $[0, 100]$ .
- The  $4 \times 1$  vector **w** containing the weights  $w_{A8}$ ,  $w_{A9}$ ,  $w_{IE}$  and  $w_{FE}$  (in that order) expressed as a fraction of unity (i.e. 0.05 for a weight of 5%).
- The scalar **X**, which is the credit received from already graded assignments and the midterm exam, expressed as a score (on the scale  $[0, 100 * (1 - w_{A8} - w_{A9} - w_{IE} - w_{FE})]$ ).

The deliverable for this question is the function described above. Please note that in this question, **grade\_o\_matic** will not provide any partial credit. Therefore, as always, you should test your function using your own data and instances. In fact, we have asked you to develop the function **r\_and\_b\_normal** in question 5 for this specific reason.

The script template for this question is provided in **A9Q6\_Template**; no modifications are required (except to remove the **\_Template** from the name). The function template is provided in **grade\_distribution\_Template.m**.