

**UNDERGRADUATE RESEARCH COURSE**  
**Mathematics Department**



Fall



Spring



Summer I



Summer II

20 21

Fox

Last Name

Brandon

First Name

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foxb3@newpaltz.edu

Email address

Course No. **MAT490**

Section # 02

CRN# 2528

Credits 3 (LA)

A. Describe the proposed research project. Type/print clearly. Use additional pages if necessary. What is the proposed method of study?

Where appropriate include a list of readings, practical experience, and/or a description of your research design.

(see attached document)

Brandon Fox 12/6/21

Signature of Student

Date

A. To be completed by instructor: On what basis will the project be evaluated? What assignments will be required?

Please see the attached

Signature of Instructor

Low

Please PRINT instructor's name:

Jenniffer Park

D. J. Hahly

Signature of Department Chair

12/14/21

Date

## "Research in Mathematics": Research Project Description

**Name:** Brandon Fox

**Date:** December 3rd, 2021

My research project was on image processing by utilizing the Gaussian diffusion/heat PDE with both the Dirichlet and Neumann boundary conditions. An essential skill to this was knowing how to do MATLAB coding.

Since I was new to MATLAB, I had done weekly MATLAB coding exercises. These exercises were in conjunction with working with images, in general, and image processing. Exercises had included, but were not limited to: image importing and making and formatting plots and subplots with MATLAB code, extraction of color channels, image transforming with power, exponential, logarithmic, linear, and piecewise linear functions, probability histogram analysis of light and dark pixels of particular images, basic linear and matrix transformations performed on images, denoising of an image by Gaussian filtering, one-dimensional and two-dimensional convolution, and variations of edge detecting.

Also, in order to arrive at the actual diffusion equation itself, I had utilized the idea of difference quotients. First, Euler's Method was rethought as a recursive definition, defined by our initial value and what's known as a "forwards difference quotient", since we approximate the next value of a function  $f$  for every  $\Delta x$  step. Next, other difference quotients had to be defined. They are obviously called the "backwards difference quotient" and the "centered difference quotient". Since the "forwards difference quotient" and the "backwards difference quotient" are defined by Taylor series, they were used in terms of their respective Taylor series to give the actual diffusion equation itself. Also, other concepts and principles (i.e. the main "Stability



Criterion",  $s = t/x^2 < 1/2$ , the "Maximum Principle") were learned and applied to enhance the idea of how the diffusion equation can be better applied, overall.

Afterwards, to gain an explicit idea of how the diffusion equation works as an application of denoising, I had done one-dimensional denoising of lines over  $t$  time. Of course, they were done under the restraints of the Dirichlet and Neumann boundary conditions. These types of boundary conditions are established under our initial data that also consists of general  $t$  time, such that these boundary conditions act as noise vanishing lines when a certain  $t$  time gets met. By doing this, we would be able to see the evolution of the denoising of the one-dimensional lines as we reach certain points in time, such that we would end up reaching an appropriate solution to the diffusion PDE when  $t$  is some time value greater than 0 and  $0 < x < L$ , where  $L$  is a Dirichlet and Neumann boundary limit, retained under our initial data;  $u(x, 0) = f(x)$ , where  $f$  is a general solution to the diffusion/heat problem, depending on what the Dirichlet and Neumann boundary conditions actually are.

Currently, I'm at the point where I hope to apply a two dimensional coefficient of diffusivity in a gradient ( $\nabla$ , or "del") direction towards sharpening the edges of some image. Hopefully, in terms of this and the diffusion equation where denoising occurs, I will arrive at the goal of creating purely denoised images without any blurring. This would well reflect the idea of "image processing", which is to improve the quality of images as effectively as possible.



**Readings:**

1. *An Image Processing Tour of Mathematics*, by Y. V. Galperin, CRC Press, first edition, pages 1-336, 2021.
2. *Introduction to MATLAB for Engineering Students*, by D. Houcque, *Northwestern University*, version 1.2, pages 1-64, 2005.
3. *Partial Differential Equations: An Introduction*, by W.A. Strauss, John Wiley & Sons Inc., second edition, pages 1-454. 2008.
4. *Perona-Malik equation and its numerical properties*, by M. Weilgus, *Uniwersytet Warszawski*, pages 1-26, 2014.



■ Assessment Criteria

Objectives	Outstanding (5)	Good (4)	Average (3)	Deficient (2)	Inadequate (1)
<b>Planning and organization</b>	Student is an active participant in the various stages of developing and implementing project and comes to meetings prepared for collaboration with mentor	Given initial encouragement, completes tasks efficiently and appropriately	Collaborates with faculty support & encouragement, completes most tasks adequately	Passive involvement, requires frequent support & reminders, tasks incomplete	Minimal involvement despite encouragement & reminders, tasks incomplete
<b>Deadlines</b>	Student consistently meets mutually negotiated deadlines and completes assigned tasks in an efficient and timely manner	Given initial encouragement, completes tasks efficiently and appropriately	Collaborates with faculty support & encouragement, completes most tasks adequately	Passive involvement, requires frequent support & reminders, tasks incomplete	Minimal involvement despite encouragement & reminders, tasks incomplete
<b>Initiative</b>	Student consistently takes the initiative to review current literature, develop plan, and propose solutions to problems as they arise	Given initial encouragement, completes tasks efficiently and appropriately	Collaborates with faculty support & encouragement, completes most tasks adequately	Passive involvement, requires frequent support & reminders, tasks incomplete	Minimal involvement despite encouragement & reminders, tasks incomplete
<b>Quality of Work</b>	Work consistently demonstrates effort, critical thought, and a developmentally appropriate	Given initial guidance and feedback, completes tasks efficiently and appropriately	Collaborates with faculty support & encouragement, completes most tasks adequately	Passive involvement, requires frequent support & reminders tasks incomplete	Minimal involvement despite encouragement & reminders, tasks incomplete



# UNDERGRADUATE RESEARCH COURSE FALL 2021 Mathematics Department

Student : Brandon Fox  
Instructor : Jeungeun Park

	synthesis of information			
<b>Personal Interaction</b>	Student exhibits a positive attitude and works well with faculty mentor. Takes leadership role when appropriate, but also demonstrates a willingness to take direction from others.	Given initial encouragement, completes tasks efficiently and appropriately	Collaborates with faculty support & encouragement, completes most tasks adequately	Passive involvement, requires frequent support & reminders, tasks incomplete  Minimal involvement despite encouragement & reminders, tasks incomplete

Scale: A 22-25, B 19-22, C 15-18, D <15

## Required assignments

1. Regular attendance.
2. Read reading material before meetings and discuss in meetings
  - a. Reference 1: Houcque, David. "Introduction to Matlab for engineering students." *Northwestern University* 1 (2005).
  - b. Reference 2: Galperin, Yevgeniy V. *An Image Processing Tour of College Mathematics*. CRC Press, 2021.
  - c. Reference 3: Strauss, Walter A. *Partial differential equations: An introduction*. John Wiley & Sons, 2007. Reference 4: Wielgus.
  - d. Reference 4: Maciek. "Perona-Malik equation and its numerical properties." *arXiv preprint arXiv:1412.6291* (2014).
3. Learn MATLAB as an introduction to programming through Reference 1 by following the guidance.
4. Learn week, learn a practice introduction to image processing with MATLAB through Reference 2. Try practice problems related to MATLAB coding.
5. Learn some basic a type of linear partial differential equations, especially heat equations, through Reference 3.
6. Learn the basic numerical methods to solve differential equations: Euler's method, finite difference method. Solve differential equations numerically by using the methods.
7. Understand how to apply a heat equation to denosing of an image and recognize an arising issue through Reference 4.

# Research in Mathematics, MAT490

Name: Brandon Fox

Date: December 14, 2021

**Problem 1.** . Consider the diffusion equation

$$u_t = u_{xx}, \quad 0 \leq x \leq 1, \quad t \geq 0, \quad (1)$$

with the Dirichlet boundary condition (2) and initial data (3)

$$u(0, t) = 0 \quad \text{and} \quad u(1, t) = 0, \quad t \geq 0, \quad (2)$$

$$u(x, 0) = f(x), \quad 0 \leq x \leq 1. \quad (3)$$

A. The initial condition is given as

$$f(x) = \begin{cases} 0, & x < 0.4, \quad x > 0.6 \\ 5, & \text{otherwise.} \end{cases} \quad (4)$$

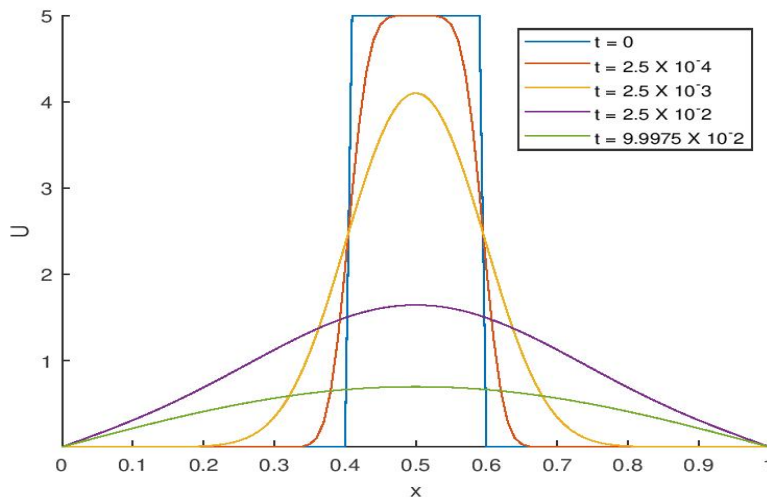


Figure 1: One Dimensional Diffusion with the Dirichlet Boundary Condition and Rectangular Initial Data Over Snapshots of  $t$  Time.

```

1  L = 1;
2  T = 0.1;
3  dx = 0.01;
4  dt = 0.25 * dx * dx;
5  s = (dt/(dx * dx));
6  J = L/dx;
7  K = T/dt;
8  x = 0:dx:L;
9  t = 0:dt:T;
10 U = ones(J + 1, K + 1);
11 for j = 1:J + 1
12     U(j,1) = initial(x(j));
13 end
14 for k = 1:K + 1
15     U(1,k) = 0;
16     U(J + 1,k) = 0;
17 end
18 for k = 1:K
19     for j = 2:J
20         U(j,k + 1) = s * (U(j + 1, k) + U(j - 1, k)) + (1 - 2*s) * U(j,k);
21     end
22 end
23 size(x)
24 %1 row and 101 columns
25 size(t)
26 %1 row and 400001 columns
27 size(U)
28 %101 rows and 400001 columns
29 %1D Dirichlet Diffusion In 2D Space
30 hold on
31 plot(x,U(:,1));
32 plot(x,U(:,11));
33 plot(x,U(:,101));
34 plot(x,U(:,1001));
35 plot(x,U(:,K));
36 hold off
37 xticks([0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1]);
38 xlabel('x');
39 yticks([1 2 3 4 5]);
40 ylabel('U');
41 legend('t = 0', 't = 2.5 X 10^-4', 't = 2.5 X 10^-3', 't = 2.5 X 10^-2', 't = 9.9975 X ...
    10^-2');

```

```

1  function f = initial(x)
2  f = zeros(size(x));
3  idx = (0.4 < x) && (x < 0.6);
4  f(idx) = 5;
5  end

```

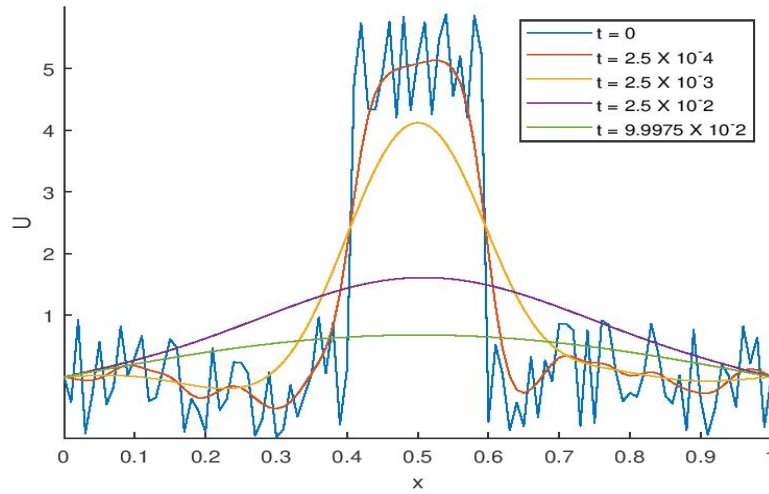


Figure 2: One Dimensional Diffusion With the Dirichlet Boundary Condition and Rectangular Initial Data Over Snapshots of  $t$  Time as an Application for Denoising.

```

1  L = 1;
2  T = 0.1;
3  dx = 0.01;
4  dt = 0.25 * dx * dx;
5  s = (dt/(dx * dx));
6  J = L/dx;
7  K = T/dt;
8  x = 0:dx:L;
9  t = 0:dt:T;
10 U = ones(J + 1, K + 1);
11 for j = 1:J + 1
12   U(j,1) = initial.noise(x(j));
13 end
14 for k = 1:K + 1
15   U(1,k) = 0;
16   U(J + 1,k) = 0;
17 end
18 for k = 1:K
19   for j = 2:J
20     U(j,k + 1) = s * (U(j + 1, k) + U(j - 1, k)) + (1 - 2*s) * U(j,k);
21   end
22 end
23 size(x)
24 %1 row and 101 columns
25 size(t)
26 %1 row and 400001 columns
27 size(U)
28 %101 rows and 400001 columns

```

```

29 %1D Dirichlet Diffusion In 2D Space
30 hold on
31 plot(x,U(:,1));
32 plot(x,U(:,11));
33 plot(x,U(:,101));
34 plot(x,U(:,1001));
35 plot(x,U(:,K));
36 hold off
37 xticks([0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1]);
38 xlabel('x');
39 yticks([1 2 3 4 5]);
40 ylabel('U');
41 legend('t = 0', 't = 2.5 X 10^-4', 't = 2.5 X 10^-3', 't = 2.5 X 10^-2', 't =
42 9.9975 X 10^-2');

```

```

1 function f = initial_noise(x)
2 f = zeros(size(x)) + (1 - 2 * (rand(1)));
3 idx = (0.4 < x) && (x < 0.6);
4 f(idx) = 5 + (1 - 2 * (rand(1)));
5 end

```

B. The initial condition is given as

$$f(x) = \begin{cases} x, & x \leq 0.5 \\ 1 - x, & \text{otherwise.} \end{cases} \quad (5)$$

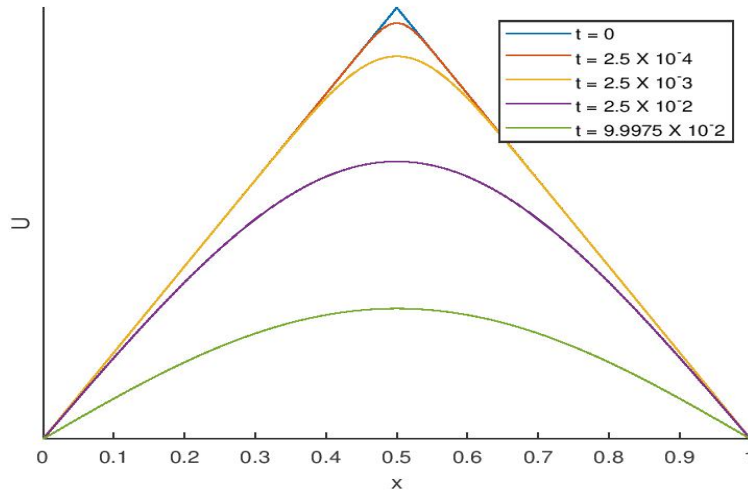


Figure 3: One Dimensional Diffusion With the Dirichlet Boundary Condition and Triangular Initial Data Over Snapshots of  $t$  Time.

```

1  L = 1;
2  T = 0.1;
3  dx = 0.01;
4  dt = 0.25 * dx * dx;
5  s = (dt/(dx * dx));
6  J = L/dx;
7  K = T/dt;
8  x = 0:dx:L;
9  t = 0:dt:T;
10 U = ones(J + 1, K + 1);
11 for j = 1:J + 1
12     U(j,1) = initial_triangle(x(j));
13 end
14 for k = 1:K + 1
15     U(1,k) = 0;
16     U(J + 1,k) = 0;
17 end
18 for k = 1:K
19     for j = 2:J
20         U(j,k + 1) = s * (U(j + 1, k) + U(j - 1, k)) + (1 - 2*s) * U(j,k);
21     end
22 end
23 size(x)
24 %1 row and 101 columns
25 size(t)
26 %1 row and 400001 columns
27 size(U)
28 %101 rows and 400001 columns
29 %1D Dirichlet Diffusion In 2D Space
30 hold on
31 plot(x,U(:,1));
32 plot(x,U(:,11));
33 plot(x,U(:,101));
34 plot(x,U(:,1001));
35 plot(x,U(:,K));
36 hold off
37 xticks([0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1]);
38 xlabel('x');
39 yticks([1 2 3 4 5]);
40 ylabel('U');
41 legend('t = 0', 't = 2.5 X 10^-4', 't = 2.5 X 10^-3', 't = 2.5 X 10^-2', 't =
42     9.9975 X 10^-2');

```

```

1  function f = initial_triangle(x)
2  f = zeros(size(x));
3  idx = (0.5 < x);
4  f(idx)=1-x;
5  idx = (x<0.5);
6  f(idx)=x;
7  idx = (x==0.5);
8  f(idx)=0.5;

```

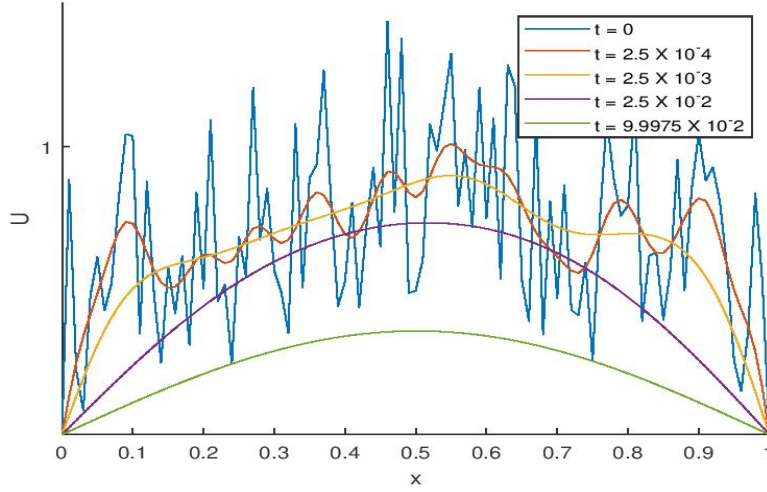


Figure 4: One Dimensional Diffusion With the Dirichlet Boundary Condition and Triangular Initial Data Over Snapshots of  $t$  Time As An Application for Denoising.

```

1 L = 1;
2 T = 0.1;
3 dx = 0.01;
4 dt = 0.25 * dx * dx;
5 s = (dt / (dx * dx));
6 J = L / dx;
7 K = T / dt;
8 x = 0:dx:L;
9 t = 0:dt:T;
10 U = ones(J + 1, K + 1);
11 for j = 1:J + 1
12     U(j,1) = initial_triangle_noise(x(j));
13 end
14 for k = 1:K + 1
15     U(1,k) = 0;
16     U(J + 1,k) = 0;
17 end
18 for k = 1:K
19     for j = 2:J
20         U(j,k + 1) = s * (U(j + 1, k) + U(j - 1, k)) + (1 - 2*s) * U(j,k);
21     end
22 end
23 size(x)
24 %1 row and 101 columns
25 size(t)
26 %1 row and 400001 columns
27 size(U)
28 %101 rows and 400001 columns

```

```

29 %1D Dirichlet Diffusion In 2D Space
30 hold on
31 plot(x,U(:,1));
32 plot(x,U(:,11));
33 plot(x,U(:,101));
34 plot(x,U(:,1001));
35 plot(x,U(:,K));
36 hold off
37 xticks([0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1]);
38 xlabel('x');
39 yticks([1 2 3 4 5]);
40 ylabel('U');
41 legend('t = 0', 't = 2.5 X 10^-4', 't = 2.5 X 10^-3', 't = 2.5 X 10^-2', 't = 9.9975 X 10^-2');

```

```

1 function f = initial_triangle_noise(x)
2 f = zeros(size(x));
3 idx = (0.5 < x);
4 f(idx) = 1 - x + rand(1);
5 idx = (x < 0.5);
6 f(idx) = x + rand(1);
7 idx = (x == 0.5);
8 f(idx) = 0.5;

```

**Problem 2.** . Consider the diffusion equation

$$u_t = u_{xx}, \quad 0 \leq x \leq 1, \quad t \geq 0, \quad (6)$$

with the Neumann boundary condition (7) and initial data (8)

$$u_x(0, t) = 0 \quad \text{and} \quad u_x(1, t) = 0, \quad t \geq 0, \quad (7)$$

$$u(x, 0) = f(x), \quad 0 \leq x \leq 1. \quad (8)$$

A. The initial condition is given as

$$f(x) = \begin{cases} 0, & x < 0.4, \quad x > 0.6 \\ 5, & \text{otherwise.} \end{cases} \quad (9)$$



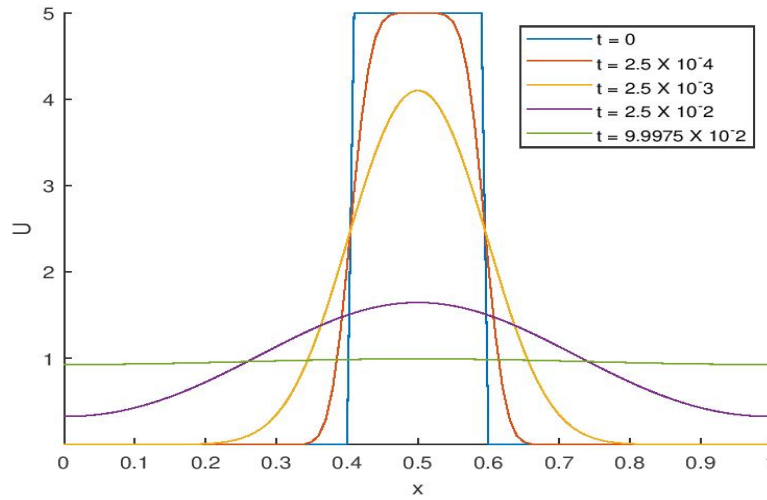


Figure 5: One Dimensional Diffusion With the Neumann Boundary Condition and Rectangular Initial Data Over Snapshots of  $t$  Time.

```

1  L = 1;
2  T = 0.1;
3  dx = 0.01;
4  dt = 0.25 * dx * dx;
5  s = (dt/(dx * dx));
6  J = L/dx;
7  K = T/dt;
8  x = 0:dx:L;
9  t = 0:dt:T;
10 U = ones(J + 1, K + 1);
11 for j = 1:J + 1
12     U(j,1) = initial(x(j));
13 end
14 for k = 1:K
15     for j = 2:J
16         U(j,k + 1) = s * (U(j + 1, k) + U(j - 1, k)) + (1 - 2*s) * U(j,k);
17     end
18     U(1,k + 1) = U(2, k + 1);
19     U(J + 1,k + 1) = U(J, k + 1);
20 end
21 size(x)
22 %1 row and 101 columns
23 size(t)
24 %1 row and 400001 columns
25 size(U)
26 %101 rows and 400001 columns
27 %1D Neumann Diffusion In 2D Space
28 hold on

```

```

29 plot(x,U(:,1));
30 plot(x,U(:,11));
31 plot(x,U(:,101));
32 plot(x,U(:,1001));
33 plot(x,U(:,K));
34 hold off
35 xticks([0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1]);
36 xlabel('x');
37 yticks([1 2 3 4 5]);
38 ylabel('U');
39 legend('t = 0', 't = 2.5 X 10^-4', 't = 2.5 X 10^-3', 't = 2.5 X 10^-2', 't =
40 9.9975 X 10^-2');

```

```

1 function f = initial(x)
2 f = zeros(size(x));
3 idx = (0.4 < x) && (x < 0.6);
4 f(idx) = 5;
5 end

```

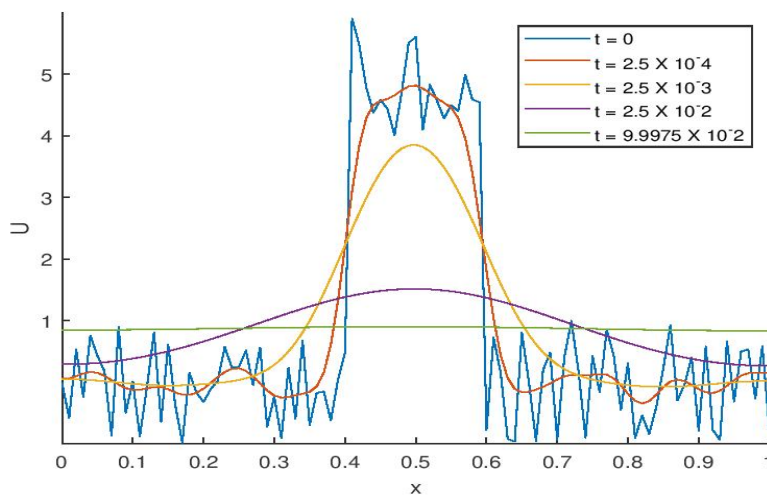


Figure 6: One Dimensional Diffusion With the Neumann Boundary Condition and Rectangular Initial Data Over Snapshots of  $t$  Time as an Application for Denoising.

```

1 L = 1;
2 T = 0.1;
3 dx = 0.01;
4 dt = 0.25 * dx * dx;
5 s = (dt/(dx * dx));
6 J = L/dx;
7 K = T/dt;
8 x = 0:dx:L;
9 t = 0:dt:T;
10 U = ones(J + 1, K + 1);
11 for j = 1:J + 1
12     U(j,1) = initial_noise(x(j));
13 end
14 for k = 1:K
15     for j = 2:J
16         U(j,k + 1) = s * (U(j + 1, k) + U(j - 1, k)) + (1 - 2*s) * U(j,k);
17     end
18     U(1,k + 1) = U(2, k + 1);
19     U(J + 1,k + 1) = U(J, k + 1);
20 end
21 size(x)
22 %1 row and 101 columns
23 size(t)
24 %1 row and 400001 columns
25 size(U)
26 %101 rows and 400001 columns
27 %1D Neumann Diffusion In 2D Space
28 hold on
29 plot(x,U(:,1));
30 plot(x,U(:,11));
31 plot(x,U(:,101));
32 plot(x,U(:,1001));
33 plot(x,U(:,K));
34 hold off
35 xticks([0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1]);
36 xlabel('x');
37 yticks([1 2 3 4 5]);
38 ylabel('U');
39 legend('t = 0', 't = 2.5 X 10^-4', 't = 2.5 X 10^-3', 't = 2.5 X 10^-2', 't =
40     9.9975 X 10^-2');

```

```

1 function f = initial_noise(x)
2 f = zeros(size(x)) + (1 - 2 * (rand(1)));
3 idx = (0.4 < x) && (x < 0.6);
4 f(idx) = 5 + (1 - 2 * (rand(1)));
5 end

```

B. The initial condition is given as

$$f(x) = \begin{cases} x, & x \leq 0.5 \\ 1 - x, & \text{otherwise.} \end{cases} \quad (10)$$

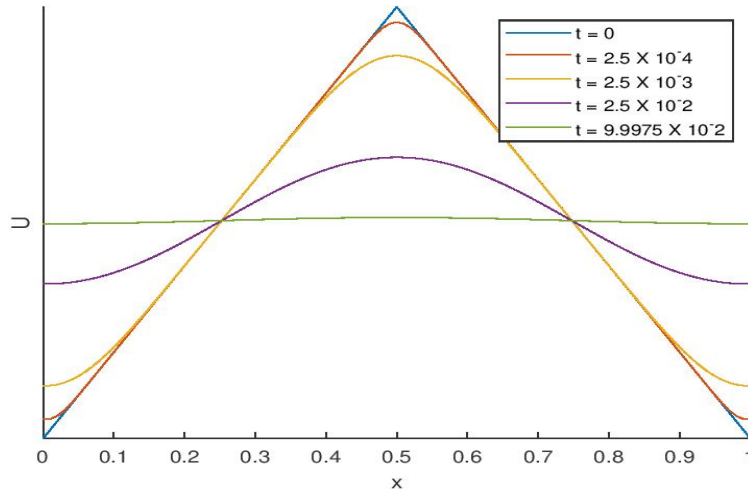


Figure 7: One Dimensional Diffusion With the Neumann Boundary Condition and Triangular Initial Data Over Snapshots of  $t$  Time.

```

1  L = 1;
2  T = 0.1;
3  dx = 0.01;
4  dt = 0.25 * dx * dx;
5  s = (dt/(dx * dx));
6  J = L/dx;
7  K = T/dt;
8  x = 0:dx:L;
9  t = 0:dt:T;
10 U = ones(J + 1, K + 1);
11 for j = 1:J + 1
12   U(j,1) = initial_triangle(x(j));
13 end
14 for k = 1:K
15   for j = 2:J
16    U(j,k + 1) = s * (U(j + 1, k) + U(j - 1, k)) + (1 - 2*s) * U(j,k);
17   end
18   U(1,k + 1) = U(2, k + 1);
19   U(J + 1,k + 1) = U(J, k + 1);
20 end
21 size(x)
22 %1 row and 101 columns
23 size(t)
24 %1 row and 400001 columns
25 size(U)
26 %101 rows and 400001 columns
27 %1D Neumann Diffusion In 2D Space
28 hold on

```

```

29 plot(x,U(:,1));
30 plot(x,U(:,11));
31 plot(x,U(:,101));
32 plot(x,U(:,1001));
33 plot(x,U(:,K));
34 hold off
35 xticks([0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1]);
36 xlabel('x');
37 yticks([1 2 3 4 5]);
38 ylabel('U');
39 legend('t = 0', 't = 2.5 X 10^-4', 't = 2.5 X 10^-3', 't = 2.5 X 10^-2', 't =
40 9.9975 X 10^-2');

```

```

1 function f = initial_triangle(x)
2 f = zeros(size(x));
3 idx = (0.5 < x);
4 f(idx)=1-x;
5 idx = (x<0.5);
6 f(idx)=x;
7 idx = (x==0.5);
8 f(idx)=0.5;

```

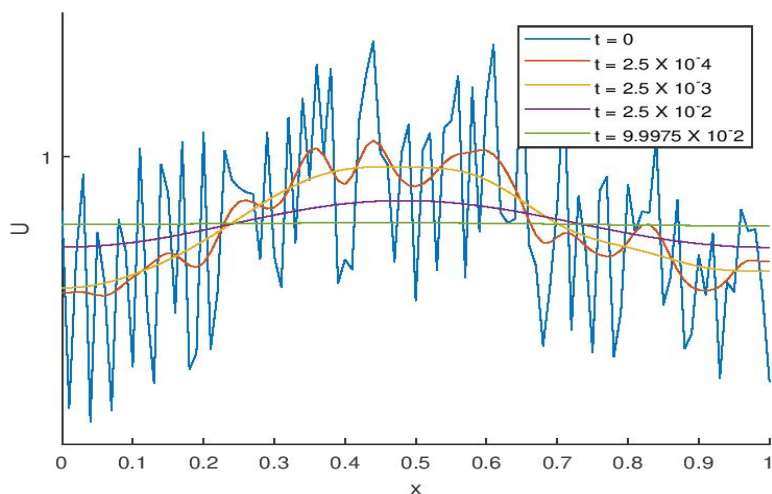


Figure 8: One Dimensional Diffusion With the Neumann Boundary Condition and Triangular Initial Data Over Snapshots of  $t$  Time as an Application for Denoising.

```

1 L = 1;
2 T = 0.1;
3 dx = 0.01;
4 dt = 0.25 * dx * dx;

```

```

5 s = (dt/(dx * dx));
6 J = L/dx;
7 K = T/dt;
8 x = 0:dx:L;
9 t = 0:dt:T;
10 U = ones(J + 1, K + 1);
11 for j = 1:J + 1
12     U(j,1) = initial_triangle_noise(x(j));
13 end
14 for k = 1:K
15     for j = 2:J
16         U(j,k + 1) = s * (U(j + 1, k) + U(j - 1, k)) + (1 - 2*s) * U(j,k);
17     end
18     U(1,k + 1) = U(2, k + 1);
19     U(J + 1,k + 1) = U(J, k + 1);
20 end
21 size(x)
22 %1 row and 101 columns
23 size(t)
24 %1 row and 400001 columns
25 size(U)
26 %101 rows and 400001 columns
27 %1D Neumann Diffusion In 2D Space
28 hold on
29 plot(x,U(:,1));
30 plot(x,U(:,11));
31 plot(x,U(:,101));
32 plot(x,U(:,1001));
33 plot(x,U(:,K));
34 hold off
35 xticks([0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1]);
36 xlabel('x');
37 yticks([1 2 3 4 5]);
38 ylabel('U');
39 legend('t = 0', 't = 2.5 x 10^-4', 't = 2.5 x 10^-3', 't = 2.5 x 10^-2', 't =
40     9.9975 x 10^-2');

```

```

1 L = 1;
2 T = 0.1;
3 dx = 0.01;
4 dt = 0.25 * dx * dx;
5 s = (dt/(dx * dx));
6 J = L/dx;
7 K = T/dt;
8 x = 0:dx:L;
9 t = 0:dt:T;
10 U = ones(J + 1, K + 1);
11 for j = 1:J + 1
12     U(j,1) = initial_triangle_noise(x(j));
13 L = 1;
14 T = 0.1;
15 dx = 0.01;
16 dt = 0.25 * dx * dx;
17 s = (dt/(dx * dx));
18 J = L/dx;
19 K = T/dt;
20 x = 0:dx:L;
21 t = 0:dt:T;
22 U = ones(J + 1, K + 1);
23 for j = 1:J + 1
24     U(j,1) = initial_triangle_noise(x(j));
25 end
26 for k = 1:K

```

```

27 for j = 2:J
28     U(j,k + 1) = s * (U(j + 1, k) + U(j - 1, k)) + (1 - 2*s) * U(j,k);
29 end
30 U(1,k + 1) = U(2, k + 1);
31 U(J + 1,k + 1) = U(J, k + 1);
32 end
33 size(x)
34 %1 row and 101 columns
35 size(t)
36 %1 row and 400001 columns
37 size(U)
38 %101 rows and 400001 columns
39 %1D Neumann Diffusion In 2D Space
40 hold on
41 plot(x,U(:,1));
42 plot(x,U(:,11));
43 plot(x,U(:,101));
44 plot(x,U(:,1001));
45 plot(x,U(:,K));
46 hold off
47 xticks([0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1]);
48 xlabel('x');
49 yticks([1 2 3 4 5]);
50 ylabel('U');
51 legend('t = 0', 't = 2.5 X 10^-4', 't = 2.5 X 10^-3', 't = 2.5 X 10^-2', 't = 9.9975 X 10^-2');

```

```

1 function f = initial-triangle-noise(x)
2 f = zeros(size(x));
3 idx = (0.5 < x);
4 f(idx) = 1 - x + rand(1);
5 idx = (x < 0.5);
6 f(idx) = x + rand(1);
7 idx = (x == 0.5);
8 f(idx) = 0.5;

```

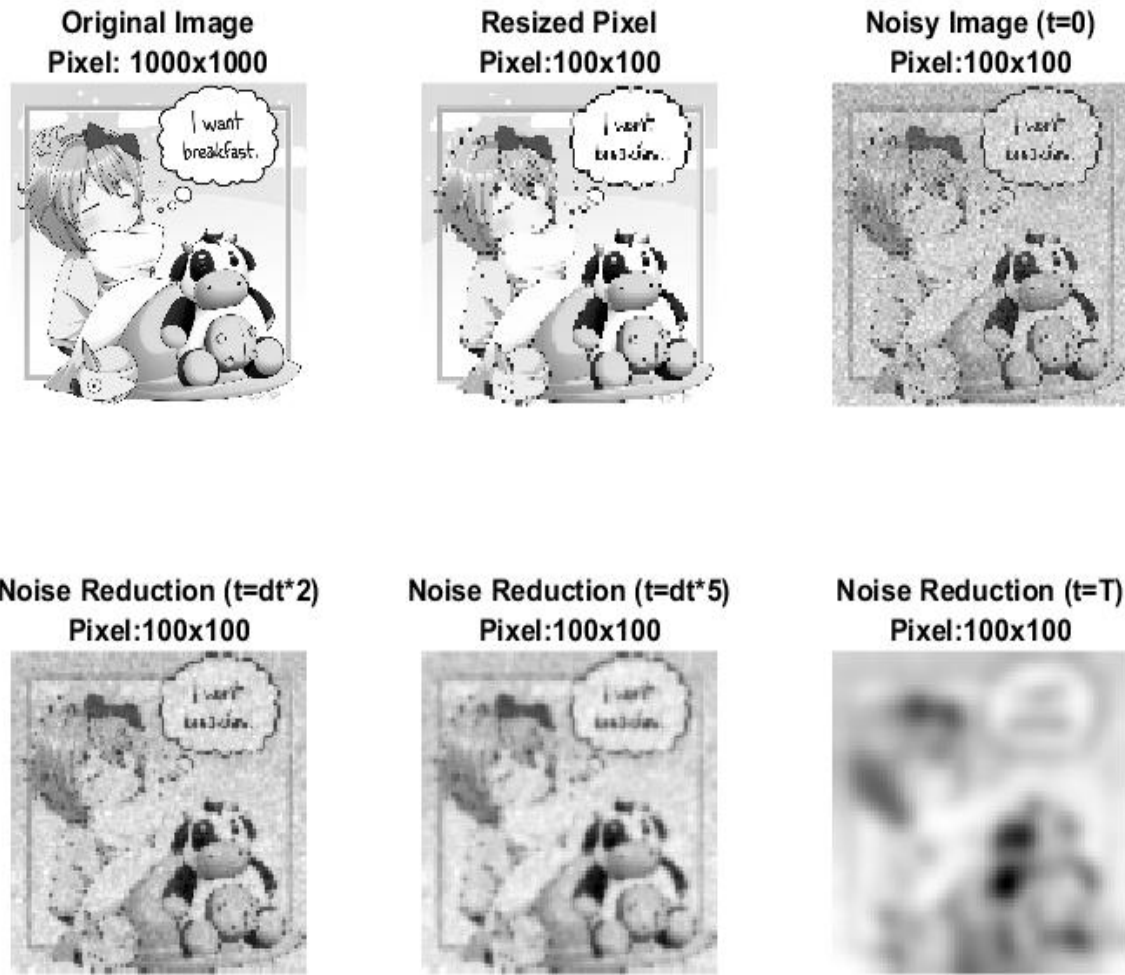


Figure 9: Two Dimensional Diffusion of A Selected Image Over  $t$  Snapshots of Time as an Application of Image Denoising. Image had to be dramatically resized from pixel size 1000 X 1000 to 100 X 100, due to memory issues appearing while executing the code. The complete noise reduced image at time  $t = T$  still requires edge detection work along the varying pixels, explaining the extreme blurriness and smoothness of the image.

```

1 Image_25 = imread('C:\Users\Brand\OneDrive\Desktop\ResearchInMathematics\Images\Image_25.jpg');
2 Image_25 = rgb2gray(Image_25);
3 Image_25=double(Image_25);
4 L = 1;
5 T = 0.001;
6 dx = 0.01;
7 dy = 0.01; %JP
8 dt = 0.25 * dx * dx *0.5; %JP
9 s = (dt/(dx * dx));
10 J1 = int32(L/dx)-1;
11 J2 = int32(L/dy)-1; %JP
12 K = int32(T/dt);
13 x = 0:dx:L;
14 y = 0:dy:L; %JP
15 t = 0:dt:T;
16 U = ones(J1 + 1, J2 +1, K + 1); %JP
17
18 % initial data for 1d
19 %for j = 1:J + 1

```



```

20 % U(j,1) = %initial-triangle-noise(x(j));
21 %end
22
23 %initial data for 2d JP
24
25 %%edited
26 % Image_25 = imread('Image-25.jpg');
27 % Image_25 = rgb2gray(Image_25);
28 % %Image_25=double(Image_25);
29 % size(Image_25);
30 % sigma = 90;
31 % Noise = normrnd(0,sigma,[1000,1000]);
32 % Image_25_Noise=Image_25+Noise;
33
34 [m,n] = size(Image_25);
35 for i = 1:10:m
36     for j = 1:10:n
37         resized_Image_25(int32(floor(i/10 +1)),int32(floor(j/10 +1)))=Image_25(i,j);
38     end
39 end
40
41 sigma = 40;
42 Noise = normrnd(0,sigma,[100,100]);
43 resized_Image_25_Noise1=resized_Image_25+Noise;
44 resized_Image_25_Noise2=(resized_Image_25)+(resized_Image_25_Noise1);
45 resized_Image_25_Noise2=double(resized_Image_25_Noise2);
46 %resized_Image_25_Noise2=padarray(resized_Image_25_Noise2,[0 0]);
47
48 initial_data = resized_Image_25_Noise2;
49
50 % v=Image_25_Noise;
51 % [m,n] = size(Image_25_Noise);
52 % for i = 1:10:m
53 %     for j = 1:10:n
54 %         initial_data(int32(floor(i/10 +1)),int32(floor(j/10 +1)))=Image_25_Noise(i,j);
55 %     end
56 % end
57 %
58 %initial_data = %reshape(new_image,100,100);
59
60 [M, N] = size(initial_data);
61 U(:, :, 1) = initial_data;
62
63 for k = 1:K
64
65     for j1 = 2:J1
66         for j2 = 2:J2
67             U(j1,j2,k + 1) = s * (U(j1 + 1, j2, k) + U(j1 - 1, j2,k)) + (1 - 2*s) * U(j1, j2,k) + s * ...
                (U(j1, j2+1, k) + U(j1, j2-1,k)) + (1 - 2*s) * U(j1, j2,k);
68         end
69     end
70
71     U(1,:, k + 1) = U(2, :, k + 1);
72     U(:, 1, k+1) = U(:, 2, k+1);
73     U(J1 + 1,:, k + 1) = U(J1, :, k + 1);
74     U(:, J2+1, k+1)=U(:, J2, k+1);
75 end
76 size(x)
77 %1 row and 101 columns
78 size(y)
79 %1 row and 101 columns
80 size(t)
81 %1 row and 400001 columns
82 size(U)
83 %101 rows and 400001 columns

```

```

84 %1D Neumann Diffusion In 2D Space
85
86 subplot(2,3,1)
87 imshow(Image_25, []);
88 title({'Original Image', 'Pixel: 1000x1000'}, 'FontSize', 7)
89
90 subplot(2,3,2)
91 imshow(initial_data, []);
92 title({'Resized Pixel', 'Pixel:100x100'}, 'FontSize', 7)
93
94 subplot(2,3,3)
95 u1 = U(:, :, 1);
96 U1 = reshape(u1, M, N);
97 imshow(U1, []);
98 title({'Noisy Image (t=0)', 'Pixel:100x100'}, 'FontSize', 7)
99
100 subplot(2,3,4)
101 u2 = U(:, :, 3);
102 U2 = reshape(u2, M, N);
103 imshow(U2, []);
104 title({'Noise Reduction (t=dt*2)', 'Pixel:100x100'}, 'FontSize', 7)
105
106 subplot(2,3,5)
107 u3 = U(:, :, 6);
108 U3 = reshape(u3, M, N);
109 imshow(U3, []);
110 title({'Noise Reduction (t=dt*5)', 'Pixel:100x100'}, 'FontSize', 7)
111
112 subplot(2,3,6)
113 u4 = U(:, :, K);
114 U4 = reshape(u4, M, N);
115 imshow(U4, []);
116 title({'Noise Reduction (t=T)', 'Pixel:100x100'}, 'FontSize', 7)

```

# Project Diary

1. September 16 meeting
  - a. Regular meeting : Mondays 5:00pm
  - b. Reading materials:
    - i. R1: An image processing tour of college mathematics by Y. V. Galperin (Chap1 ~ Chap 7 (hopefully) )
    - ii. R2: PDEs by W. A. Strauss (Chap1 and Chap2 (partly) )
    - iii. Intro. MATLAB pdf file (only W1 and W2)
  - c. MATLAB
2. W1: September 20
  - a. Euler's method
  - b. MATLAB
  - c. To do:
    - i. R1 - Chapter 1. Read carefully including examples. Do MATLAB exercises. Put some exercise results into a folder 'W1'  
**\*UNABLE TO DO CH.1.3 and CH.1.4: PROBLEM 2b); IN SUMMARY, TOO MANY THINGS WENT WRONG WITH THE INSTALLATION OF THE TOOLBOXES AND IMPORTING THE IMAGES ONTO THE MATLAB ON REGULAR BROWSER IN THE VDI INTERFACE!**  
**[Okay! --Jeungeun]**
    - ii. Intro. MATLAB pdf file - p.4 ~ p.34. Just try examples.
    - iii. R2 - Read 1.1
3. W2: September 27
  - a. Review W1
  - b. To do:
    - i. R1 - 1.2 - Try Q5 again. Let me know if you have any further question.
    - ii. R1 - Chapter 2: MATLAB exercises. Please put some exercise results into W2 folder
    - iii. Intro MATLAB pdf file - p.35 ~ end. Try examples (not exercises), and read carefully since some sections do not have examples.
    - iv. R2 - Read 1.2. Please follow the examples very carefully.
4. W3: October 4
  - a. Review W2 - (J. Park: Check Section 2.6 - Q2)
  - b. To do:
    - i. R1 - Chapter 3
    - ii. R2 - Read 1.3
5. W4: October 15
  - a.
  - b. To do:
    - i. R1 - Read Chap 3 - Exercise (Jeungeun)
    - ii. R2 - Read 1.4 and 1.5
6. W5: October 18 (This upcoming Monday? Also, the 22nd is my 22nd birthday (22 on 22!)! - Brandon Fox)

- a. Chap3, Pseudo-code
  - b. To do:
    - i. R1 - Chap 4
    - ii. R2 - Read 2.3 and 2.4
  
- 7. W6: October 25
  - a. Review: Brandon's email question. Pseudo code for Euler method
  - b. To do:
    - i. R1 - Sections 5.1-5.2
    - ii. R2 - Section 8.1. Please go over Example 1 and try to digest it so we can use it for our matlab code.
  
- 8. W7: Nov 1
  - a. Review
  - b. To do:
    - i. R1 - Sections 5.3
    - ii. R2 - Chap 8
    - iii. Tomorrow: (Computational methods)MATLAB code: Euler method for  $y'=y$ ,  $y(0)=1$  .. Pseudo code.
  
- 9. W8: Nov 8
  - a. Review
  - b. To do:
    - i. R1 - take a break for this week :)
    - ii. R2 - Chap 8.2 - Boundary value problems
  
- 10. W9 - W16:
  - a. Review
  - b. To do:
    - i. R2 - Chap 8.2 - Boundary value problems - finite difference methods
    - ii. Read reading material on Perona-Malik equation
    - iii. Solve diffusion equations with MATLAB:
      - Euler method to solve  $y'=f(t,y)$
      - 1D Diffusion equation with Dirichlet or Neumann boundary condition
      - 2D Diffusion equation with Neumann boundary condition
      - Add noise in an initial datum
      - Reduce noise in an initial datum as time changes
    - iv. Write a result in LaTeX

# Example of weekly work by Brandon Fox:

## Image processing textbook Practice problem in Section 5.3

```
%1
Image_25 = imread('C:\Users\Brand\OneDrive\Desktop\ResearchInMathematics\Images\Image_25.jpg');
Image_25 = rgb2gray(Image_25);
imshow(Image_25);

%2 and 3

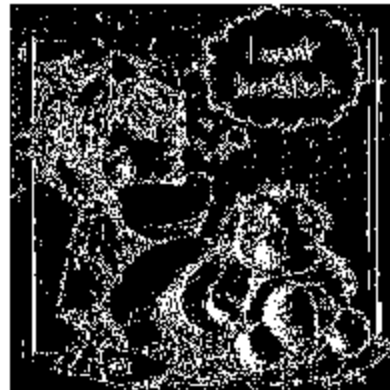
%Gradient edge dector
[M, N]=size(Image_25);
G=zeros(M,N);
Gx=G; Gy=G;
for m=1:M-1
for n=1:N-1
Gx(m,n)=Image_25(m,n+1)-Image_25(m,n);
Gy(m,n)=Image_25(m+1,n)-Image_25(m,n);
G(m,n)=(sqrt(Gx(m,n))) & (2+Gy(m,n)) & (2);
end
end
subplot(1,2,1); imshow(Image_25); title('Original Image');
subplot(1,2,2); imshow(abs(G),[]); title('Gradient Edge Detector Image');
```

---

Original Image



Gradient Edge Detector Image



```
%Roberts edge detector
[M, N]=size(Image_25);
G=zeros(M,N);
G_1=G; G_2=G;
for m=1:M-1
for n=1:N-1
G_1(m,n)=Image_25(m+1,n+1)-Image_25(m,n);
G_2(m,n)=Image_25(m+1,n)-Image_25(m,n+1);
G(m,n)=(sqrt(G_1(m,n)) & (2+G_2(m,n)) & (2));
end
end
subplot(1,2,1); imshow(Image_25); title('Original Image');
subplot(1,2,2); imshow(abs(G),[]); title('Roberts Edge Detector Image');
```

---

Original Image



Roberts Edge Detector Image



```
%Prewitt edge detector
f1 = (double(Image_25));
f2 = (double(Image_25));

h = fspecial('Prewitt');
P=uint8(round(filter2(h,Image_25)));

subplot(1,2,1); imshow(Image_25); title('Original Image');
subplot(1,2,2); imshow(P); title('Prewitt Edge Detector Image');
```

---

Original Image



Prewitt Edge Detector Image



```
%Sobel edge detector
I = double(Image_25);
for i=1:size(I,1)-2
    for j=1:size(I,2)-2
        G_1=((2*I(i+2,j+1)+I(i+2,j)+I(i+2,j+2))-(2*I(i,j+1)+I(i,j)+I(i,j+2)));
        G_2=((2*I(i+1,j+2)+I(i,j+2)+I(i+2,j+2))-(2*I(i+1,j)+I(i,j)+I(i+2,j)));
        G(i,j)=sqrt(G_1.^2+G_2.^2);
    end
end
subplot(1,2,1); imshow(Image_25); title('Original Image');
subplot(1,2,2); imshow(abs(G),[]); title('Sobel Edge Detector Image');
```



---

Original Image



Sobel Edge Detector Image



```
%Laplacian edge detector
L=[-1 -1 -1;-1 8 -1; -1 -1 -1];
G=uint8(filter2(L,Image_25,'same'));
subplot(1,2,1); imshow(Image_25); title('Original Image');
subplot(1,2,2); imshow(abs(G),[]); title('Laplacian Edge Detector Image');
```

---

**Original Image**



**Laplacian Edge Detector Image**



```
%5
size(Image_25);
Image_25=double(Image_25);
sigma = 40;
Noise = normrnd(0,sigma,[1000,1000]);
Image_25_Noise=Image_25+Noise;

imshow(Image_25_Noise,[]); title('Noisy Image');
```

---

**Original Image**



**Noisy Image**



```
%Gradient edge dector
size(Image_25);
Image_25=double(Image_25);
sigma = 40;
Noise = normrnd(0,sigma,[1000,1000]);
Image_25_Noise=Image_25+Noise;
[M, N]=size(Image_25_Noise);
G=zeros(M,N);
Gx=G; Gy=G;
for m=1:M-1
for n=1:N-1
Gx(m,n)=Image_25_Noise(m,n+1)-Image_25_Noise(m,n);
Gy(m,n)=Image_25_Noise(m+1,n)-Image_25_Noise(m,n);
G(m,n)=(sqrt(Gx(m,n))) + (2+Gy(m,n)) + (2);
end
end
imshow(abs(G),[]); title('Gradient Edge Detector Image');
```

---

Original Image



Gradient Edge Detector Image



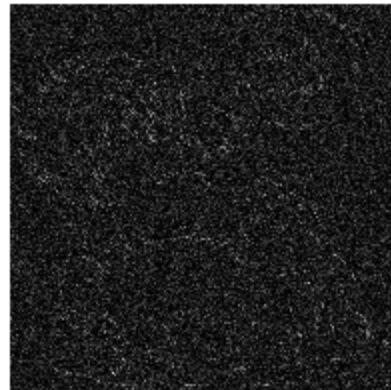
```
%Roberts edge detector
size(Image_25);
Image_25=double(Image_25);
sigma = 40;
Noise = normrnd(0,sigma,[1000,1000]);
Image_25_Noise=Image_25+Noise;
[M, N]=size(Image_25_Noise);
G=zeros(M,N);
G_1=G; G_2=G;
for m=1:M-1
for n=1:N-1
G_1(m,n)=Image_25_Noise(m+1,n+1)-Image_25_Noise(m,n);
G_2(m,n)=Image_25_Noise(m+1,n)-Image_25_Noise(m,n+1);
G(m,n)=(sqrt(G_1(m,n)) + (2+G_2(m,n)) + (2));
end
end
imshow(abs(G),[]); title('Roberts Edge Detector Image');
```

---

Original Image



Roberts Edge Detector Image



```
%Prewitt edge detector
size(Image_25);
Image_25=double(Image_25);
sigma = 40;
Noise = normrnd(0,sigma,[1000,1000]);
Image_25_Noise=Image_25+Noise;

f1 = (double(Image_25_Noise));
f2 = (double(Image_25_Noise));

h = fspecial('Prewitt');
P=uint8(round(filter2(h,Image_25_Noise)));

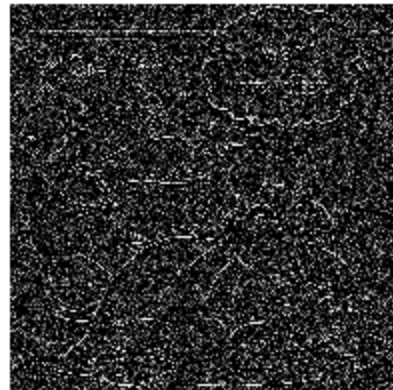
imshow(P); title('Prewitt Edge Detector Image');
```

---

Original Image



Prewitt Edge Detector Image



```
%Sobel edge detector
size(Image_25);
Image_25=double(Image_25);
sigma = 40;
Noise = normrnd(0,sigma,[1000,1000]);
Image_25_Noise=Image_25+Noise;
for i=1:size(Image_25_Noise,1)-2
    for j=1:size(Image_25_Noise,2)-2
        G_1=((2*Image_25_Noise(i+2,j+1)+Image_25_Noise(i+2,j)+Image_25_Noise(i
+2,j+2))-(2*Image_25_Noise(i,j+1)+Image_25_Noise(i,j)+Image_25_Noise(i,j+2)));
        G_2=((2*Image_25_Noise(i+1,j+2)+Image_25_Noise(i,j+2)+Image_25_Noise(i
+2,j+2))-(2*Image_25_Noise(i+1,j)+Image_25_Noise(i,j)+Image_25_Noise(i+2,j)));
        G(i,j)=sqrt(G_1.^2+G_2.^2);
    end
end
imshow(abs(G),[]); title('Sobel Edge Detector Image');
```

---

**Original Image**



**Sobel Edge Detector Image**



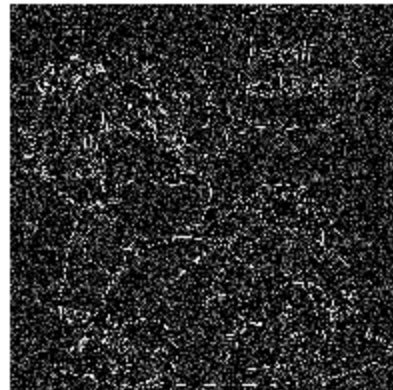
```
%Laplacian edge detector
size(Image_25);
Image_25=double(Image_25);
sigma = 10;
Noise = normrnd(0,sigma,[1000,1000]);
Image_25_Noise=Image_25+Noise;
L=[-1 -1 -1;-1 8 -1; -1 -1 -1];
G=uint8(filter2(L,Image_25_Noise,'same'));
imshow(abs(G),[]); title('Laplacian Edge Detector Image');
```

---

Original Image



Laplacian Edge Detector Image



```
%7
Image_25=double(Image_25);
Image_25 = Image_25 - min(Image_25(:));
Image_25 = Image_25 / max(Image_25(:));
noise_per=2;
delta = (noise_per*std(Image_25(:)))^2
Image_25_Noise1=normrnd(0,delta,[1000 1000]);
Image_25_Noise2=(Image_25)+(Image_25_Noise1);
Image_25_Noise2=double(Image_25_Noise2);

sigma = 2.06;
Box_Size=3;
[x,y] = meshgrid(-Box_Size:Box_Size,-Box_Size:Box_Size);
M = size(x,1)-1;
N = size(y,1)-1;

Exp = -(x.^2+y.^2)/(2*sigma.^2);
Filter = ((1)/(2*pi*sigma.^2))*exp(Exp);
```



---

```

Blurred_Image=size(Image_25_Noise2);
Image_25_Noise2 = padarray(Image_25_Noise2,[0 0]);

for i=2:size(Image_25_Noise2,1)-M
    for j=2:size(Image_25_Noise2,2)-N
        T= Image_25_Noise2(i:i+M,j:j+M).*Filter;
        Blurred_Image(i,j)=sum(T(:));
    end
end
subplot(1,3,1); imshow(Image_25); title('Original Image');
subplot(1,3,2); imshow(Image_25_Noise2); title('Noisy Image');
subplot(1,3,3); imshow(Blurred_Image); title('Blurred Image');

delta =

    0.2273

```



```
%Gradient edge detector
```

---

```

Image_25=double(Image_25);
Image_25 = Image_25 - min(Image_25(:));
Image_25 = Image_25 / max(Image_25(:));
noise_per=2;
delta = (noise_per*std(Image_25(:)))^2
Image_25_Noise1=normrnd(0,delta,[1000 1000]);
Image_25_Noise2=(Image_25)+(Image_25_Noise1);
Image_25_Noise2=double(Image_25_Noise2);

Grad_Grad = imgradient(Image_25_Noise2);

sigma = 2;
S = imgaussfilt(Image_25_Noise2, sigma);
G = edge(S, 'Gradient');

subplot(1,3,1); imshow(Image_25); title('Original Image');
subplot(1,3,2); imshow(Image_25_Noise2); title('Noisy Image');
subplot(1,3,3); imshow(abs(G),[]); title('Gradient Edge Detector Image');

```

```

delta =

    0.2273

```

```

Error using edge>parse_inputs (line 518)
Invalid input string or character vector: 'Gradient'.

```

```

Error in edge (line 213)
[a,method,thresh,sigma,thinning,H,kx,ky] = parse_inputs(args{:});

```

```

Error in Ch_5_3 (line 198)
G = edge(S, 'Gradient');

```

```

Error in evalmxdom>instrumentAndRun (line 114)
text = evalc(evalstr);

```

```

Error in evalmxdom (line 21)
[data,text,laste] =
    instrumentAndRun(file,cellBoundaries,imageDir,imagePrefix,options);

```

```

Error in publish

```

```

Error in connector.internal.fevalMatlab

```

```

Error in connector.internal.fevalJSON

```

```

Error in internal.matlab.publish.captureFigures (line 14)
drawnow;

```

```

Error in internal.matlab.publish.PublishFigures/leavingCell (line 78)
    newFigures = internal.matlab.publish.captureFigures;

```

```

Error in snapnow>leavingCell (line 208)

```

---

```

        newFiles = data.plugins(iPlugins).instance.leavingCell(iCell);

Error in snapnow (line 144)
        data = leavingCell(iCell(k), data, doCapture(k));

Error in Ch_5_3 (line 152)
Image_25=double(Image_25);

%Roberts edge detector
Image_25=double(Image_25);
Image_25 = Image_25 - min(Image_25(:));
Image_25 = Image_25 / max(Image_25(:));
noise_per=2;
delta = (noise_per*std(Image_25(:)))^2
Image_25_Noise1=normrnd(0,delta,[1000 1000]);
Image_25_Noise2=(Image_25)+(Image_25_Noise1);
Image_25_Noise2=double(Image_25_Noise2);

Roberts_Grad = imgradient(Image_25_Noise2);

sigma = 2;
S = imgaussfilt(Image_25_Noise2, sigma);
G = edge(S, 'Roberts');

subplot(1,3,1); imshow(Image_25); title('Original Image');
subplot(1,3,2); imshow(Image_25_Noise2); title('Noisy Image');
subplot(1,3,3); imshow(abs(G),[]); title('Roberts Edge Detector Image');

%Prewitt edge detector
Image_25=double(Image_25);
Image_25 = Image_25 - min(Image_25(:));
Image_25 = Image_25 / max(Image_25(:));
noise_per=2;
delta = (noise_per*std(Image_25(:)))^2
Image_25_Noise1=normrnd(0,delta,[1000 1000]);
Image_25_Noise2=(Image_25)+(Image_25_Noise1);
Image_25_Noise2=double(Image_25_Noise2);

Prewitt_Grad = imgradient(Image_25_Noise2);

sigma = 2;
S = imgaussfilt(Image_25_Noise2, sigma);
G = edge(S, 'Prewitt');

subplot(1,3,1); imshow(Image_25); title('Original Image');
subplot(1,3,2); imshow(Image_25_Noise2); title('Noisy Image');
subplot(1,3,3); imshow(abs(G),[]); title('Prewitt Edge Detector Image');

%Sobel edge detector
Image_25=double(Image_25);
Image_25 = Image_25 - min(Image_25(:));
Image_25 = Image_25 / max(Image_25(:));
noise_per=2;
delta = (noise_per*std(Image_25(:)))^2

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Image_25_Noise1=normrnd(0,delta,[1000 1000]);
Image_25_Noise2=(Image_25)+(Image_25_Noise1);
Image_25_Noise2=double(Image_25_Noise2);

Sobel_Grad = imgradient(Image_25_Noise2);
hy = -fspecial('sobel')
hx = hy';

sigma = 2;
S = imgaussfilt(Image_25_Noise2, sigma);
G = imgradient(S,'CentralDifference');

subplot(1,3,1); imshow(Image_25); title('Original Image');
subplot(1,3,2); imshow(Image_25_Noise2); title('Noisy Image');
subplot(1,3,3); imshow(abs(G),[]); title('Sobel Edge Detector Image');

%Laplacian edge detector
Image_25=double(Image_25);
Image_25 = Image_25 - min(Image_25(:));
Image_25 = Image_25 / max(Image_25(:));
noise_per=2;
delta = (noise_per*std(Image_25(:)))^2;
Image_25_Noise1=normrnd(0,delta,[1000 1000]);
Image_25_Noise2=(Image_25)+(Image_25_Noise1);
Image_25_Noise2=double(Image_25_Noise2);

Laplacian_Grad = imgradient(Image_25_Noise2);

sigma = 2;
S = imgaussfilt(Image_25_Noise2, sigma);
G = edge(S,'log');

subplot(1,3,1); imshow(Image_25); title('Original Image');
subplot(1,3,2); imshow(Image_25_Noise2); title('Noisy Image');
subplot(1,3,3); imshow(abs(G),[]); title('Laplacian Edge Detector Image');

%8
Binary_Image_25 = Image_25 > 0.5;

subplot(1,2,1); imshow(Image_25); title("Original Image");
subplot(1,2,2); imshow(Binary_Image_25); title("Binary Image");

%10

%a
Binary_Image_25 = Image_25 > 0.5;
S_E=[0 0 0; 0 1 0; 0 0 0];
L1 = imdilate(Binary_Image_25,S_E);

subplot(1,2,1); imshow(Binary_Image_25); title('Binary Image');
subplot(1,2,2); imshow(L1); title('Linear Edge Dilated Image #1');

Binary_Image_25 = Image_25 > 0.5;
S_E=[0 0 0; 1 1 1; 0 0 0];
m=length(Binary_Image_25);

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n=length(S_E);
N=max(m,n);
Binary_Image_25=[Binary_Image_25,zeros(3,N-m)];
S_E=[S_E,zeros(3,N-n)];
for n=1:N
    Y(n)=0;
    for i=1:N
        j=n-i+1;
        if(j<=0)
            j=N+j;
        end
        Y(n)=[Y(n)+Binary_Image_25(i)*S_E(j)];
    end
end
subplot(1,2,1); imshow(Binary_Image_25); title('Binary Image');
subplot(1,2,2); imshow(Y); title('Circular Edge Dilated Image #1');

%b
Binary_Image_25 = Image_25 > 0.5;
S_E =[0 0 0; 1 1 1; 0 0 0];
L2 = imdilate(Binary_Image_25,S_E);

subplot(1,2,1); imshow(Binary_Image_25); title('Binary Image');
subplot(1,2,2); imshow(L2); title('Linear Edge Dilated Image #2');

Binary_Image_25 = Image_25 > 0.5;
S_E =[0 0 0; 1 1 1; 0 0 0];
m=length(Binary_Image_25);
n=length(S_E);
N=max(m,n);
Binary_Image_25=[Binary_Image_25,zeros(3,N-m)];
S_E=[S_E,zeros(3,N-n)];
for n=1:N
    Y(n)=0;
    for i=1:N
        j=n-i+1;
        if(j<=0)
            j=N+j;
        end
        Y(n)=[Y(n)+Binary_Image_25(i)*S_E(j)];
    end
end
subplot(1,2,1); imshow(Binary_Image_25); title('Binary Image');
subplot(1,2,2); imshow(Y); title('Circular Edge Dilated Image #2');

%c
Binary_Image_25 = Image_25 > 0.5;
S_E =[0 1 0; 0 1 0; 0 1 0];
L3 = imdilate(Binary_Image_25,S_E);

subplot(1,2,1); imshow(Binary_Image_25); title('Binary Image');
subplot(1,2,2); imshow(L3); title('Linear Edge Dilated Image #3');

Binary_Image_25 = Image_25 > 0.5;
S_E =[0 1 0; 0 1 0; 0 1 0];

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```

m=length(Binary_Image_25);
n=length(S_E);
N=max(m,n);
Binary_Image_25=[Binary_Image_25,zeros(3,N-m)];
S_E=[S_E,zeros(3,N-n)];
for n=1:N
    Y(n)=0;
    for i=1:N
        j=n-i+1;
        if(j<=0)
            j=N+j;
        end
        Y(n)=[Y(n)+Binary_Image_25(i)*S_E(j)];
    end
end
subplot(1,2,1); imshow(Binary_Image_25); title('Binary Image');
subplot(1,2,2); imshow(Y); title('Circular Edge Dilated Image #3');

%d
Binary_Image_25 = Image_25 > 0.5;
S_E =[0 1 0; 1 1 1; 0 1 0];
L4 = imdilate(Binary_Image_25,S_E);

subplot(1,2,1); imshow(Binary_Image_25); title('Binary Image');
subplot(1,2,2); imshow(L4); title('Linear Edge Dilated Image #4');

Binary_Image_25 = Image_25 > 0.5;
S_E =[0 1 0; 1 1 1; 0 1 0];
m=length(Binary_Image_25);
n=length(S_E);
N=max(m,n);
Binary_Image_25=[Binary_Image_25,zeros(3,N-m)];
S_E=[S_E,zeros(3,N-n)];
for n=1:N
    Y(n)=0;
    for i=1:N
        j=n-i+1;
        if(j<=0)
            j=N+j;
        end
        Y(n)=[Y(n)+Binary_Image_25(i)*S_E(j)];
    end
end
subplot(1,2,1); imshow(Binary_Image_25); title('Binary Image');
subplot(1,2,2); imshow(Y); title('Circular Edge Dilated Image #4');

%e
Binary_Image_25 = Image_25 > 0.5;
S_E = [1 0 1; 0 1 0; 1 0 1];
L5 = imdilate(Binary_Image_25,S_E);

subplot(1,2,1); imshow(Binary_Image_25); title('Binary Image');
subplot(1,2,2); imshow(L5); title('Linear Edge Dilated Image #5');

Binary_Image_25 = Image_25 > 0.5;

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```

S_E = [1 0 1; 0 1 0; 1 0 1];
m=length(Binary_Image_25);
n=length(S_E);
N=max(m,n);
Binary_Image_25=[Binary_Image_25,zeros(3,N-m)];
S_E=[S_E,zeros(3,N-n)];
for n=1:N
    Y(n)=0;
    for i=1:N
        j=n-i+1;
        if(j<=0)
            j=N+j;
        end
        Y(n)=[Y(n)+Binary_Image_25(i)*S_E(j)];
    end
end
subplot(1,2,1); imshow(Binary_Image_25); title('Binary Image');
subplot(1,2,2); imshow(Y); title('Circular Edge Dilated Image #5');

%f
Binary_Image_25 = Image_25 > 0.5;
S_E = ones(3,3);
L6 = imdilate(Binary_Image_25,S_E);

subplot(1,2,1); imshow(Binary_Image_25); title('Binary Image');
subplot(1,2,2); imshow(L6); title('Linear Edge Dilated Image #6');

Binary_Image_25 = Image_25 > 0.5;
S_E = ones(3,3);
m=length(Binary_Image_25);
n=length(S_E);
N=max(m,n);
Binary_Image_25=[Binary_Image_25,zeros(3,N-m)];
S_E=[S_E,zeros(3,N-n)];
for n=1:N
    Y(n)=0;
    for i=1:N
        j=n-i+1;
        if(j<=0)
            j=N+j;
        end
        Y(n)=[Y(n)+Binary_Image_25(i)*S_E(j)];
    end
end
subplot(1,2,1); imshow(Binary_Image_25); title('Binary Image');
subplot(1,2,2); imshow(Y); title('Circular Edge Dilated Image #6');

```

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