

Survey of Scientific Computing (SciComp 301)

Dave Biersach
Brookhaven National
Laboratory
dbiersach@bnl.gov

Exam 3
Total of 100 points

1. Predator-Prey Modelling

https://en.wikipedia.org/wiki/Lotka%E2%80%93Volterra_equations

In the **q01** folder, edit the C++ CERN ROOT application to calculate the **Lotka-Volterra** (1920) differential equations for given characteristics & initial conditions



$$\alpha = 2, \beta = 1.1, \gamma = 1.0, \delta = 0.9$$

 $x(0) = 1, y(0) = 0.5$

In this model, at time t: x(t) represents the **prey** population y(t) represents **predator** population



Fig. 1.2 – Vito Volterra

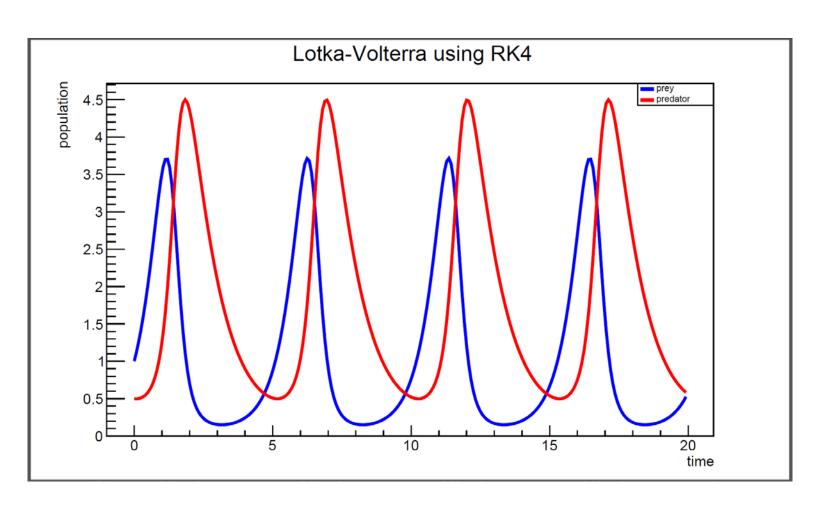
Their system of *coupled* non-linear first order differential equations will be solved using the **4**th **order Runge-Kutta** method

1. Predator-Prey Modelling

```
29
       // Lotka-Volterra {Prey} dx/dt
       double d prey(double x, double y, double t)
30
31
                                                                          d_prey() = \frac{dx}{dt}
                                                    You must write
32
           return 0;
33
                                                      this function
34
35
       // Lotka-Volterra {Predator} dy/dt
36
       double d predator(double x, double y, double t)
37
     \square{
38
           return 0;
                                                                          d_predator() = \frac{dy}{dt}
                                                    You must write
39
                                                      this function
40
41
       int main()
42
     \square{
43
           // Initial time
44
           double t = 0.0;
45
                                                           Provide these
46
           // Initial prey population %
                                                               values
47
           double x = 0.0;
48
49
           // Initial predator population %
50
           double y = 0.0;
51
```

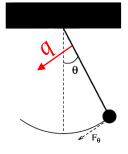
1. Predator-Prey Modelling

Expected Output (Approved Solution)



2. Damped Pendulum

In the **q02** folder, edit the C++ CERN ROOT application to accurately model a pendulum damped with a frictional resistance **q** directly *proportional* to its **angular velocity**



Referring to Session 19 Lab 04, we must introduce an additional resistive force term into the equation of motion

$$\frac{d^2\theta}{dt^2} = -\frac{g}{l}\theta - \frac{q}{q}\frac{d\theta}{dt}$$
 Dampening force constant $\frac{q}{q}$

$$\frac{d\omega}{dt} = -\frac{g}{l}\theta - \frac{d\theta}{dt}$$

$$\omega_{i+1} = \omega_i - \frac{g}{l}\theta_i\Delta t - q\omega_i\Delta t$$

$$\frac{d\theta}{dt} = \omega$$

$$\theta_{i+1} = \theta_i + \omega_{i+1}\Delta t$$

2. Damped Pendulum

Assume a damping factor q = 1

$$\omega_{i+1} = \omega_i - \frac{g}{l}\theta_i \Delta t - \mathbf{q}\omega_i \Delta t$$

Add damping term

const double phaseConstant = g / length;

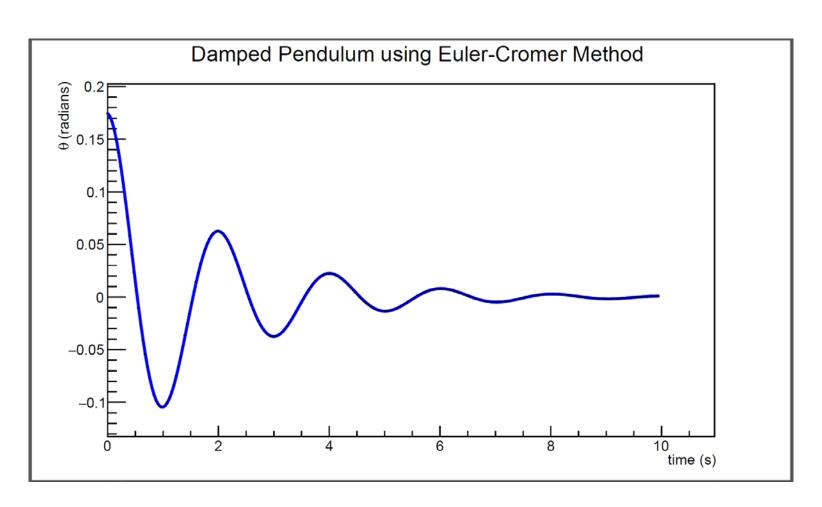
```
// Perform Euler method to estimate differential equation
for (int step{}; step < timeSteps - 1; ++step) {
   omega[step + 1] = omega[step] - phaseConstant * theta[step] * deltaTime;
   theta[step + 1] = theta[step] + omega[step] * deltaTime;
   timeAt[step + 1] = timeAt[step] + deltaTime;
}
</pre>
```

$$\theta_{i+1} = \theta_i + \omega_{i+1} \Delta t$$

Insert Cromer's correction

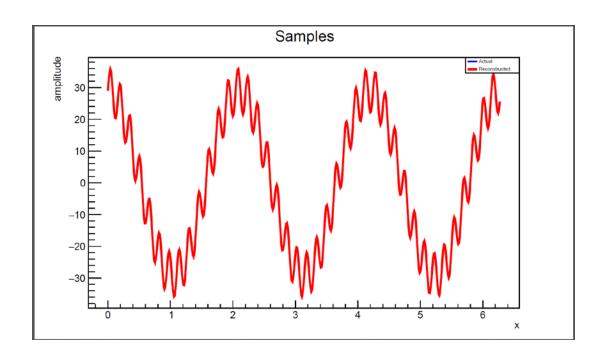
2. Damped Pendulum

Expected Output (Approved Solution)



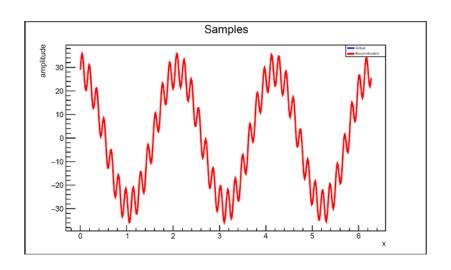
3. High Frequency Filter

In the **q03** folder, edit the C++ CERN ROOT application to filter out the high frequency noise embedded in a signal using the methods learned in Session 21

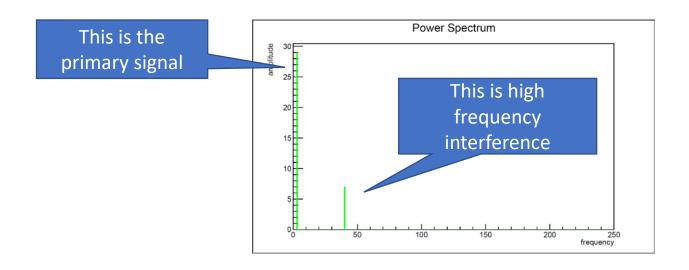


High frequency interference is distorting the capture of a clean primary signal. We want to remove this interference when reconstructing the signal using the inverse discrete Fourier transform (IDFT)

3. High Frequency Filter



The DFT identifies the constituent simple waves composing the complex wave. By ignoring high frequency simple waves when reconstructing the signal using the IDFT, the interference can be removed



3. High Frequency Filter

back into the

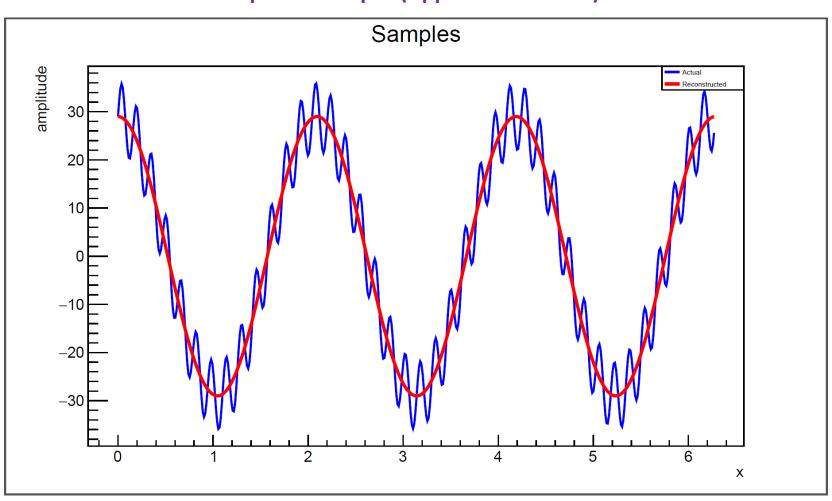
reconstructed signal

```
The term # is the
                                                                              frequency of each
                          void CalcIDFT()
                   66
                   67
                        □{
                                                                           successive simple wave
                              size t sample count{ yAct.size() };
                   68
                   69
                              size t term count{ fCos.size() };
                   70
                              for (size t i{}; i < sample count; ++i)</pre>
                   71
                   72
                   73
                                  double xs = xRad.at(i);
                   74
                                  double yt{};
                                   for (size t term{}; term < term count; ++term)</pre>
                   75
                   76
                   77
                                       vt += fCos.at(term) * cos(term * xs);
                   78
                                       yt += fSin.at(term) * sin(term * xs);
                                  yEst.push_back(yt);
Insert logic to not add
high frequency waves
```

10

3. High Frequency Filter

Expected Output (Approved Solution)



4. Newtonian Kinematics

In the **q04** folder, edit the C++ console application to determine the initial velocity v_0 and constant acceleration a for a particle obeying these distance measures over the stated time. Assume SI units.

time (s)	distance (m)
0.0000	0.0000
1.0000	29.1199
2.0000	83.5010
3.0000	163.1435
4.0000	268.0472
5.0000	398.2123
6.0000	553.6386
7.0000	734.3263
8.0000	940.2752
9.0000	1,171.4855
10.0000	1,427.9570

Using the method of least squares, fit an appropriate equation from kinematics that governs the behavior of this particle

4. Newtonian Kinematics

```
Enter given data

x = time, y = distance

int main()

double vecX[10] { 1,2,3,4,5,6,7,8,9,10 };

double vecY[10] { 1,2,3,4,5,6,7,8,9,10 };
```

Edit code to display correct values for acceleration and initial velocity

5. Combustion of Octane

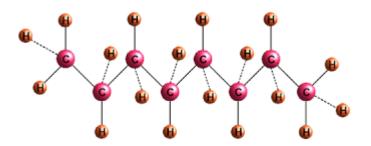
In the **q05** folder, edit file **octane.txt** to correctly balance the combustion reaction equation of **gasoline**

Ensure the application emits the correct molar ratios

Refer to Session 17 for assistance on how to encode a chemical equation into the expected input file format

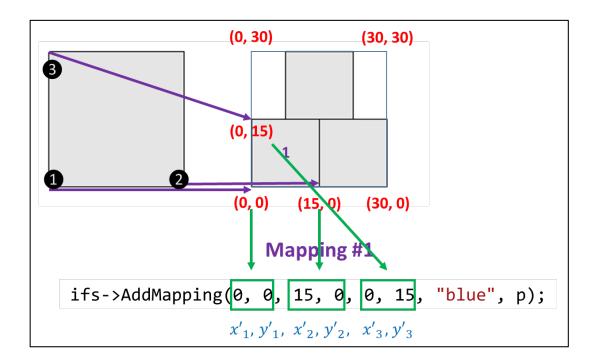
Combustion

HydroCarbon Carbon Dioxide
$$CH + O_2 \longrightarrow CO_2 + H_2O$$
Oxygen Water



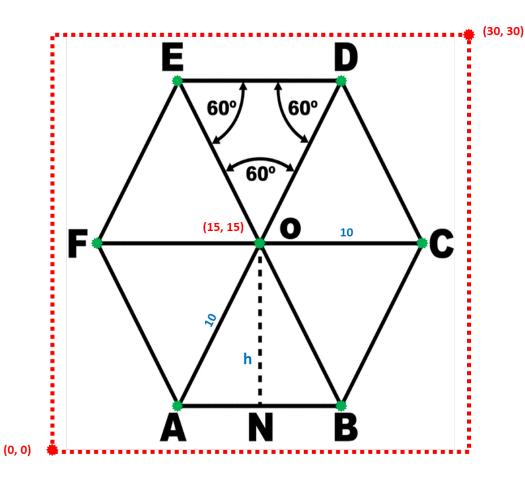
6. Hexagonal Fractal

In the q06 folder, edit the C++ Allegro application to draw a hexagonal fractal using an Iterated Function System



Provide the necessary coordinates to create six affine transforms (mappings) that cover a regular hexagon with side length 10

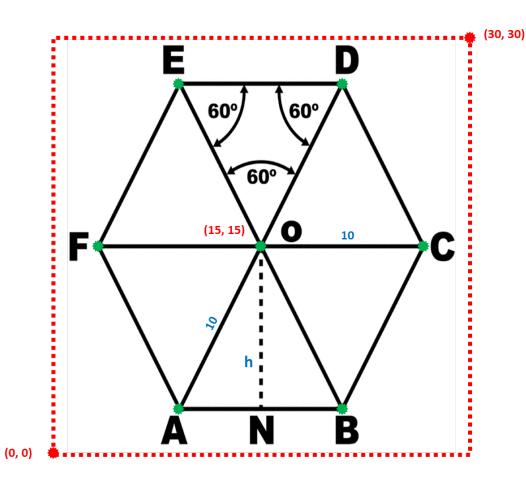
Refer to **Session 24** for assistance on how to encode mappings



The IFS base frame is a square measuring (0,0) - (30,30)

The hexagon is centered on point (15, 15)

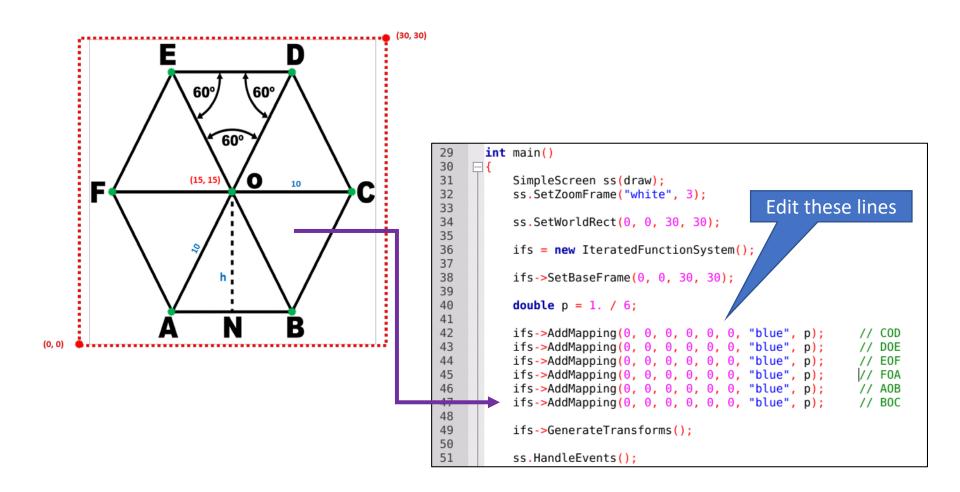
The hexagon has side length of 10

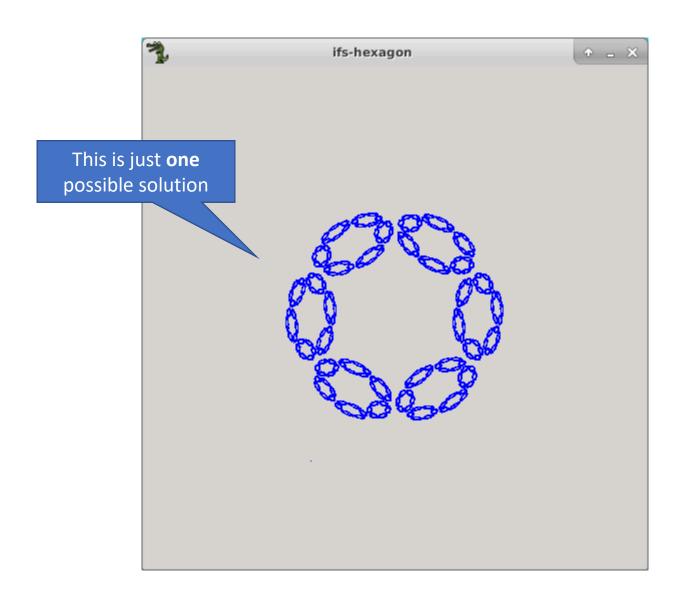


Find the Cartesian coordinates for points A, B, C, D, E, F, O

Encode these six mappings:

- 1. COD
- 2. DOE
- 3. EOF
- 4. FOA
- 5. AOB
- 6. BOC

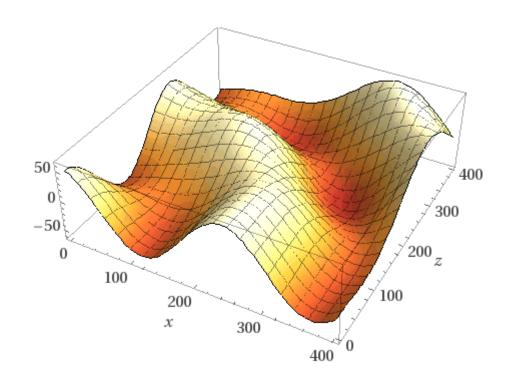




7. Surface Interpolation

In the **q07** folder, edit the C++ Allegro application to determine the optimal IDW **power** that minimizes the RMSD of this model

$$y = -15\sin\left(\frac{x}{40}\right)\cos\left(\frac{z}{40}\right) + 50\cos\left(\frac{\sqrt{x^2 + z^2}}{40}\right)$$



7. Surface Interpolation

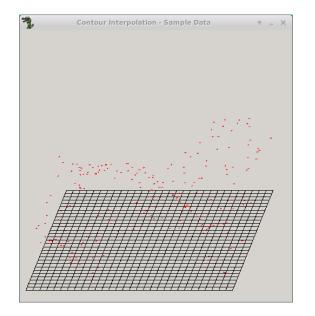
```
double GetActHeight(double x, double z)

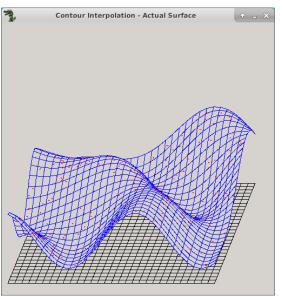
double GetActHeight(double x, double z)

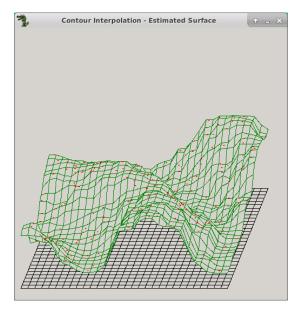
return 0;

Edit this function
}
```

```
File Edit View Terminal Tabs Help
Press S to see only sample data
Press A to see actual ocean floor
Press E to see estimated ocean floor
Press - to reduce p by 0.1
Press + to increase p by 0.1
```

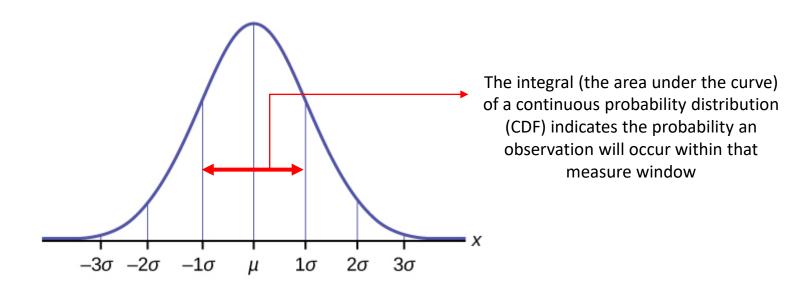






8. Standard Normal Monte Carlo

In the **q08** folder, edit the C++ Allegro application to use the **Monte Carlo** method to estimate the probability that a normally distributed random variable will fall within \pm the first standard deviation (σ) of its mean (μ)



Assume we have a **standard normal** distribution for this problem!

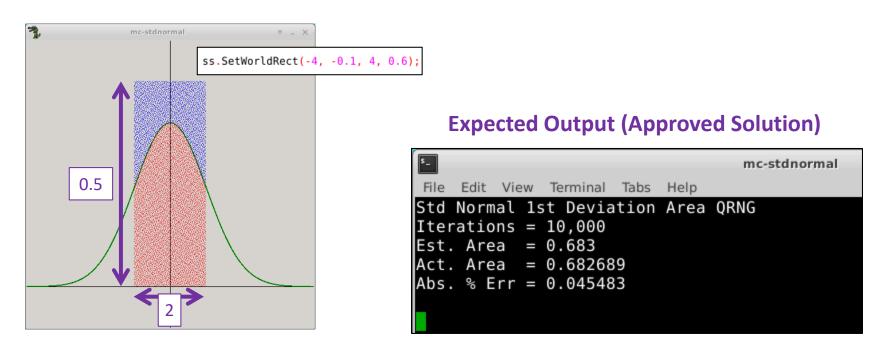
8. Standard Normal Monte Carlo

We will use the **Niederreiter** QRNG

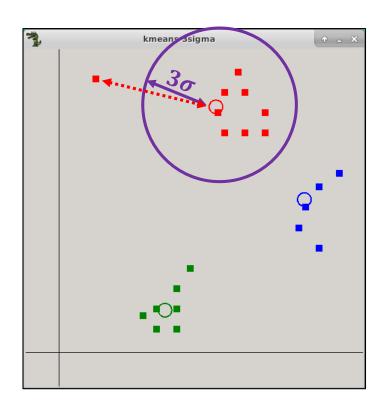
```
// mc-stdnormal.cpp
                                        Implement the function
 3
       #include "stdafx.h"
                                        for a standard normal
      #include "SimpleScreen.h"
       #include "Niederreiter2.h"
                                                  CDF
 6
       using namespace std;
 8
       double f(double x)
10
11
           return 0;
                                                       for (int i{}; i < iterations; ++i)</pre>
                                            38
12
                                            39
                                            40
                                                            grng.Next(2, &seed, r);
                                            41
                                                            double x = r[0] * -2.0 - 1.0;
                                            42
                                                            double y = r[1] * -0.5;
                                            43
                                                            if (true)
         Edit this logic to only
                                                                ss.DrawPoint(x, y, "red");
                                            45
         count points that are
                                            46
                                                                count++;
          under the curve f(x)
                                            47
                                            48
                                                            else
                                            49
                                                                ss.DrawPoint(x, y, "blue");
                                            50
```

8. Standard Normal Monte Carlo

```
double estArea = (double)count / iterations;
                         Insert the actual
                                                                   double actArea = 1;
                                                                   double err = (actArea - estArea) / actArea * 100;
                                                      56
                         (expected) value
                                                      57
                            for this area
                                                      58
                                                                   cout << "Std Normal 1st Deviation Area ORNG" << endl</pre>
                                                      59
                                                                         << "Iterations = " << iterations << endl</pre>
                                                      60
                                                                         << "Est. Area = " << estArea << endl</pre>
                                                                         << "Act. Area = " << actArea << endl</pre>
                                                      61
                                                                         << "Abs. % Err = " << abs(err) << endl << endl:</pre>
                                                      62
\frac{dots_{inside}}{dots_{inside}} = \frac{area_{under\ curve}}{dots_{under\ curve}}
                                                      63
                                                      64
dots_{total}
                area<sub>rectangle</sub>
```



9. kMeans Eviction Criteria



$$\sigma = \sqrt{\sigma^2} = \sqrt{\frac{\sum (x_i - \mu)^2}{N}}$$

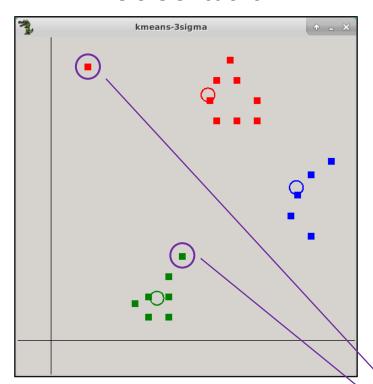
In the **q09** folder, edit the C++ Allegro application to determine the *reasonableness* of evicting data outliers that are beyond three sigma's (3σ) distance from the centroid {mean μ point} of all the kMeans clusters

If the points are distributed **normally** around the cluster's mean point, then 99.97% of all points should fall within 3σ distance of the cluster centroid

What are the possible complications with using 3σ as a measure of cluster inclusion?

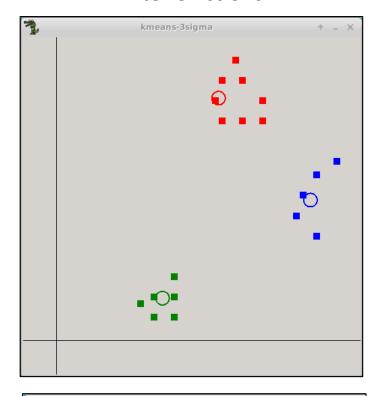
9. kMeans Eviction Criteria

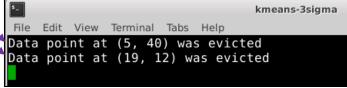
Before evictions



Did we evict too many data points? Does the algorithm converge?

After evictions



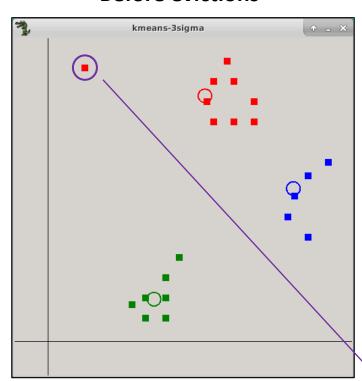


9. kMeans Eviction Criteria

```
// Phase III
232
233
            // Evict any data point more than 4 sigmas away
            // from the mean of its currently assigned cluster
234
235
            if (converged)
236
237
                for (auto c : *clusters)
238
239
                    // Calculate the standard deviation of distance from
240
                    // each cluster's center to each of its assigned points
241
                    c->dist sigmas = 0;
242
                    int count = 0;
243
                    for (auto dp : *dataPoints)
244
245
                         if (dp->c == c)
246
247
                             c\rightarrow dist sigmas += sqrt(pow(dp\rightarrow x - c\rightarrow x, 2) +
                                                                                     This is the 3 in the
248
                                                     pow(dp->y - c->y, 2));
249
                             count++;
                                                                                          3\sigma eviction
250
251
                                                                                            threshold
252
                    c->dist sigmas = sgrt(c->dist sigmas / count) * 3;
253
254
                    // For each data point belonging to this cluster,
255
                    // check if its distance is > 4 sigmas from the center
256
                    for (auto dp : *dataPoints)
257
258
                         if (dp->c == c)
259
260
                             double dist = sqrt(pow(dp->x - c->x, 2) +
261
                                                 pow(dp->y - c->y, 2));
262
                             if (dist > c->dist sigmas)
263
264
                                 // We need to evict this data point from its cluster
                                 // so the algorithm clearly has not yet converged
265
                                 converged = false;
266
```

9. kMeans Eviction Criteria

Before evictions



Which values of n in the cutoff metric of $n\sigma$ will only prune likely true outliers?

After evictions

