

Survey of Scientific Computing (SciComp 301)

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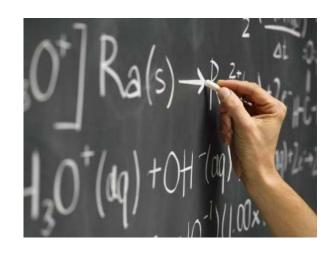
Session 17
Computational Chemistry,
Clustering

Session Goals

- Balance ionic chemical equations using linear programming
- Correlate "principle of atom conservation" (POAC) to the minimization of an objective function
- Represent molar & charge ratios as linear constraints
- Encode chemical equations as a matrix within a text file
- Implement brute-force searching using a stack data structure instead of using recursion or nested for() loops

Balancing Ionic Equations

 Can we write a program to balance ionic equations?



```
5 Types of Reactions

C + O_2 \rightarrow CO_2

H_2O_2 \rightarrow O_2 + H_2O

Zn + CuCl_2 \rightarrow ZnCl_2 + Cu

KI + Pb(NO_3)_2 \rightarrow Pbl_2 + KNO_3

CH_4 + O_2 \rightarrow H_2O + CO_2
```

Balancing Ionic Equations

Step 1: Write reactants and products

$$C_3H_8 + O_2 \rightarrow CO_2 + H_2O_3$$

Step 2: Find one atom that occurs only in one substance on both sides

$$C_3H_8 + O_2 \rightarrow CO_2 + H_2$$

Step 3: Find coeffici

Step only one substance occui

$$1C_3H_8 + O_2 \rightarrow 3CO_2 + H_2O$$

Step 5: Find coefficient to balance s atom

1: Write reactants and products

$$C_3H_8 + O_2 \rightarrow CO_2 + H_2O$$

2: Find one atom that occurs only in one ance on both sides

 $C_3H_8 + O_2 \rightarrow CO_2 + H_2O$

3: Find coefficients are to the smallest whole number values

 $C_3H_8 + O_2 \rightarrow CO_2 + H_2O$

3: Find coefficients are to the smallest whole number values

 $C_3H_8 + SO_2 \rightarrow CO_2 + 4H_2O$

Step 7: Make sure the coefficients are to the smallest whole number values

 $C_3H_8 + SO_2 \rightarrow CO_2 + 4H_2O$

Step 8: Check the balanced equation

Step 7: Make sure the coefficients are reduced

$$1C_3H_8 + 5O_9 \rightarrow 3CO_9 + 4H_9O$$

Step 8: Check the balanced equation

$$1C_3H_8 + 5O_2 \rightarrow 3CO_2 + 4H_2O$$

3C, 8H, 10O 3C, 10O, 8H

Chemistry is Optimization

Consider a chemical formula as an optimization problem:

$$MnO_4^- + H_2O_2 + H^+ \rightarrow Mn^{2+} + O_2 + H_2O$$

$$x_0 \text{MnO}_4^- + x_1 \text{H}_2 \text{O}_2 + x_2 \text{H}^+ \rightarrow x_3 \text{Mn}^{2+} + x_4 \text{O}_2 + x_5 \text{H}_2 \text{O}$$

Develop the objective function and the constraints:

obj: minimize
$$\sum_{i=0}^{n} x_i$$

due to Principle of Atom Conservation (POAC)

Chemistry is Optimization

$$x_0 \, \text{MnO}_4^- + x_1 \, \text{H}_2 \, \text{O}_2 + x_2 \, \text{H}^+ \rightarrow x_3 \, \text{Mn}_2^{2+} + x_4 \, \text{O}_2 + x_5 \, \text{H}_2 \, \text{O}$$

$$obj: \, minimize \, \sum_{i=0}^n x_i$$

$$x_0 = x_3 \qquad \text{To balance the Manganese (Mn)}$$

$$4x_0 + 2x_1 = 2x_4 + x_5 \qquad \text{To balance the Oxygen (O)}$$

$$2x_1 + x_2 = 2x_5 \qquad \text{To balance the Hydrogen (H)}$$

$$-x_0 + x_2 = 2x_3 \qquad \text{To balance the charges}$$

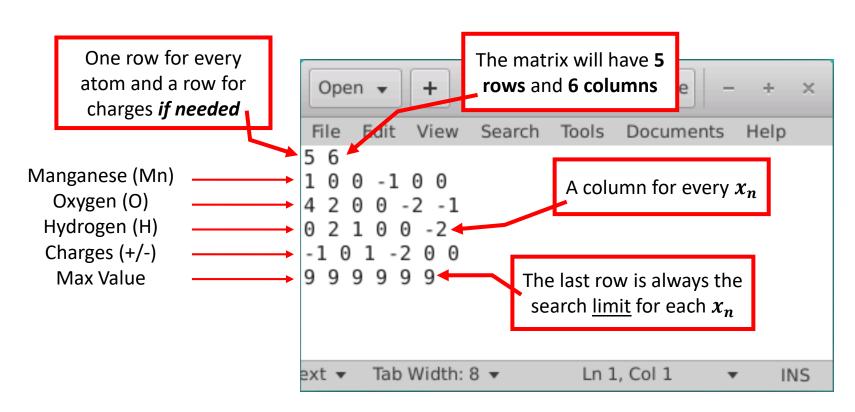
Chemistry is Optimization

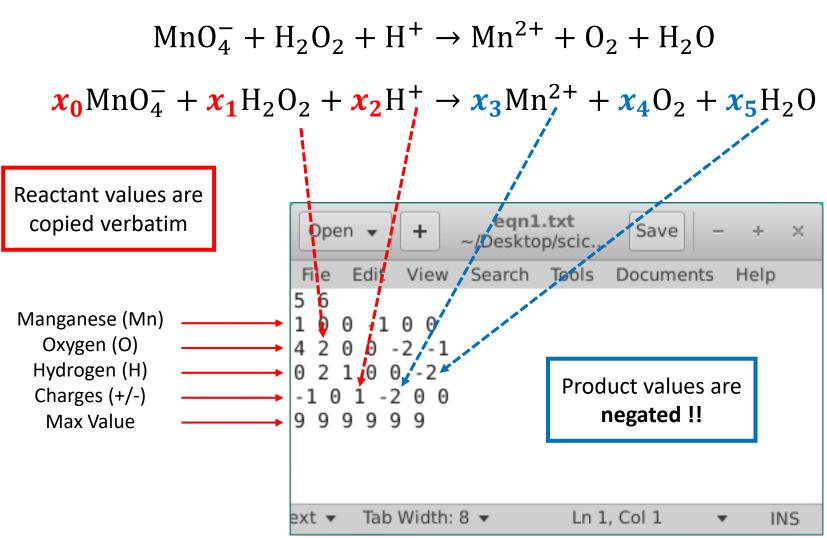
- We have transformed solving a chemical equation into a solving a linear optimization problem
 - The Simplex method is a known good approach to solve linear programming problems in reasonable time
 - However managing slack variables and the tableau algorithm can be intimidating for novice programmers
- For small problems, we can just enumerate the **search space** (defined by all **permutations** of coefficients subject to the constraints) in order to find all of the possible solutions!

- We can encode a chemical equation using a 2D matrix
 - There will be a horizontal <u>row</u> for each <u>element</u> (atom)
 - We may need another row to encode the ionic charges
 - The last row will always contain the maximum values we want to consider for each unknown coefficient
- Each molecule is a term in the equation
 - There will be a vertical <u>column</u> for each <u>molecule</u> (or charged ion)
 - Every column is another unknown we need to resolve
- The matrix elements will be stored in a text (.txt) file
 - The first line will describe the # of cols and # of rows

$$MnO_4^- + H_2O_2 + H^+ \rightarrow Mn^{2+} + O_2 + H_2O$$

$$x_0MnO_4^- + x_1H_2O_2 + x_2H^+ \rightarrow x_3Mn^{2+} + x_4O_2 + x_5H_2O$$





 A potential valid set of unknown values will yield a sum of zero when convolved with each row (constraint) in the matrix

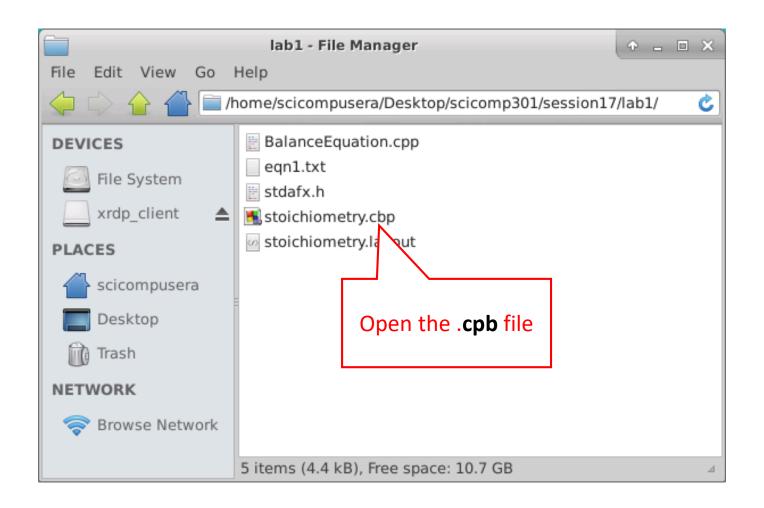
$$x_0 \text{MnO}_4^- + x_1 \text{H}_2 \text{O}_2 + x_2 \text{H}^+ \rightarrow x_3 \text{Mn}^{2+} + x_4 \text{O}_2 + x_5 \text{H}_2 \text{O}_3$$

For the Oxygen atom these coefficients are taken from 3rd row in matrix

$$(4)x_0 + (2)x_1 + (0)x_2 + (0)x_3 + (-2)x_4 + (-1)x_5 = 0$$

- The current search values for each x_n are multiplied by each column coefficient in each row in the equation matrix if the sum for every row == 0, then we have a potential solution!
- The final solution is where $\sum x_n$ has the <u>lowest</u> value (POAC)

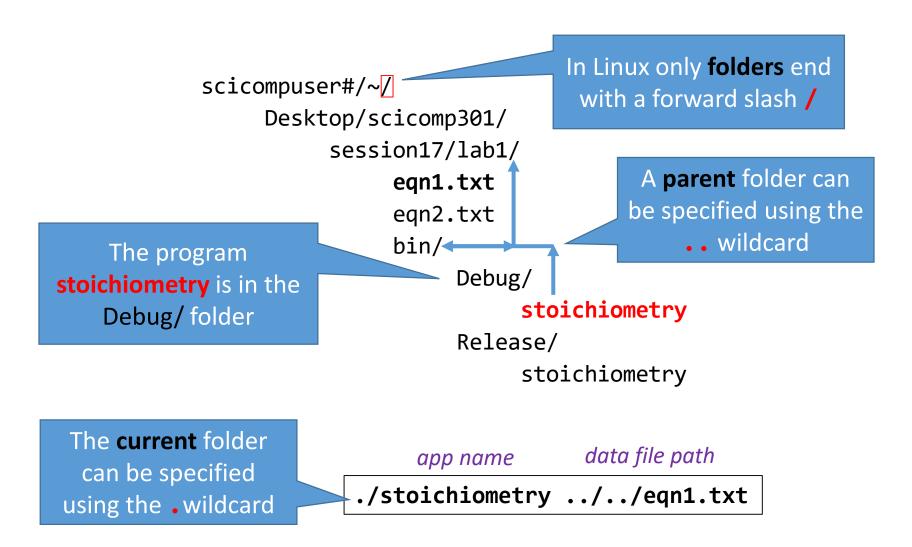
Open Lab 1 – Stoichiometry



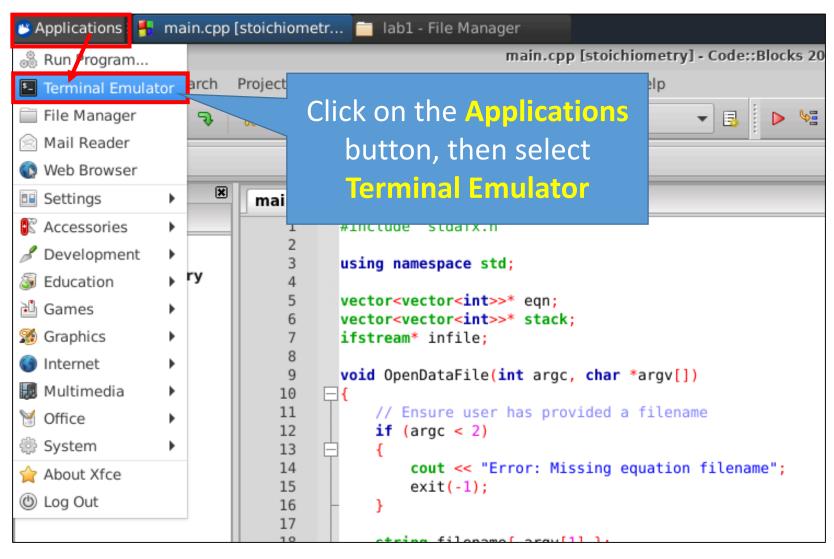
Build Lab 1 – Stoichiometry

```
BalanceEquation.cpp [stoichiometry] - Code::Blocks 16.01
               Search Project Build Debug
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                                | 🔍 🙉 🗄 🗯
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Management
                         ×
                              BalanceEquation.
Projects Symbols
                                                             47
                                 48
                                                                   1,egn->back().at(i) + 1 });
   Workspace
                                 49
   stoichiometry
                                          Build the solution
                                 50
                                 51
       Sources
                                 52
       Headers
                                 53
                                            <del>τοι (size_ι ){};} < ed</del>n->size() - 1;++i) {
                                                int sum = 0;
      Others
                                 54
                                 55
                                                for (size t i{};i < stack->size();++i)
                                                   sum += eqn->at(j).at(i) * stack->at(i).at(0);
                                 56
                                 57
                                                if (sum != 0)
                                                    return false:
                                 58
                                 59
                                            return true:
                                 60
                                 61
                                 62
                                        bool FindSolution()
                                 63
                                 64
                                            int stackLevel = 0:
                                 65
                                            while (stackLevel >= 0) {
                                 66
                                                while (stack->at(stackLevel).at(0)
                                 67
                                                   < stack->at(stackLevel).at(1))
                                 68
                                 69
                                                   if (stackLevel == (int)stack->size() - 1) break:
```

Project Folder & File Structure



Opening A Linux Terminal



Run Lab 1 – Stoichiometry



Verify Chemical Equation 1

$$\begin{array}{c} x_0 \, \text{MnO}_4^- + x_1 \, \text{H}_2 \, \text{O}_2 + x_2 \, \text{H}^+ \rightarrow x_3 \, \text{Mn}^{2+} + x_4 \, \text{O}_2 + x_5 \, \text{H}_2 \, \text{O} \\ & \text{Terminal - scicompusera@ip-172-31-71-74: ~/Desktop/scicomp301/session17/lab1/bin/Debug} \\ & \text{Scicompusera@ip-172-31-71-74:~} \, \text{cd Desktop/scicomp301/session17/lab1/bin/Debug} \\ & \text{scicompusera@ip-172-31-71-74:~/Desktop/scicomp301/session17/lab1/bin/Debug$./stoichiometry ../../eqn1.txt} \\ & \text{v0 = 2} \\ & \text{x1 = 1} \\ & \text{x2 = 6} \\ & \text{x3 = 2} \\ & \text{x4 = 3} \\ & \text{x5 = 4} \\ & \text{scicompusera@ip=172-31-71-74:~/Desktop/scicomp301/session17/lab1/bin/Debug$} \\ & \text{2MnO}_4^- + 1 \, \text{H}_2 \, \text{O}_2 + 6 \, \text{H}^+ \rightarrow 2 \, \text{Mn}^{2+} + 3 \, \text{O}_2 + 4 \, \text{H}_2 \, \text{O}} \\ & \text{2MnO}_4^- + 1 \, \text{H}_2 \, \text{O}_2 + 6 \, \text{H}^+ \rightarrow 2 \, \text{Mn}^{2+} + 3 \, \text{O}_2 + 4 \, \text{H}_2 \, \text{O}} \\ \end{array}$$

$$Cr_2O_7^{2-} + H^+ + H_2C_2O_4 \rightarrow Cr^{3+} + H_2O + CO_2$$

$$x_0 \text{Cr}_2 \text{O}_7^{2-} + x_1 \text{H}^+ + x_2 \text{H}_2 \text{C}_2 \text{O}_4 \rightarrow x_3 \text{Cr}^{3+} + x_4 \text{H}_2 \text{O} + x_5 \text{CO}_2$$

A Row for Every Element

Matrix Size (Rows=6 Cols=6)

Six Rows:

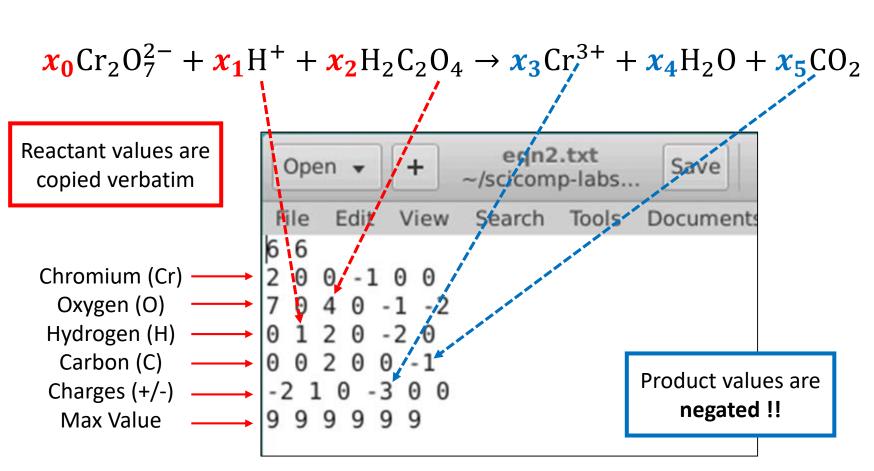
- 1. Chromium (Cr)
- 2. Oxygen (O)
- 3. Hydrogen (H)
- 4. Carbon (C)
- 5. Charges
- 6. Max Search Value per Term

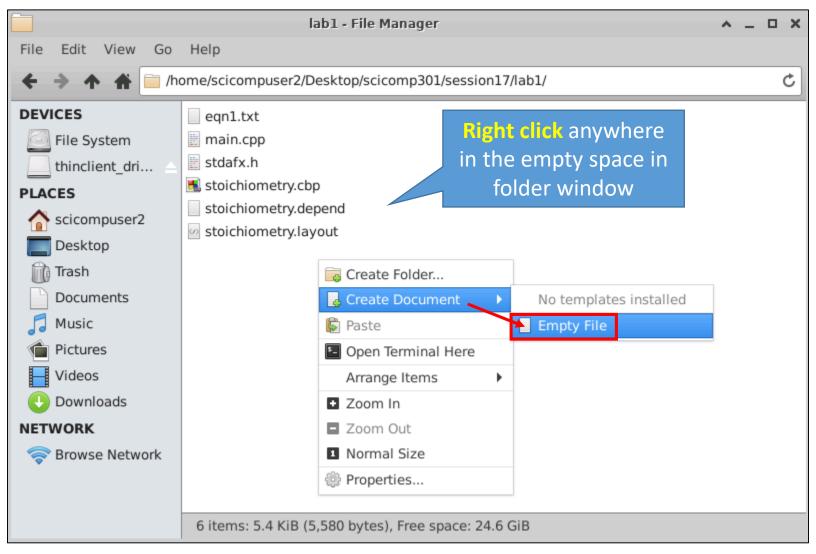
A Column for Every Molecule

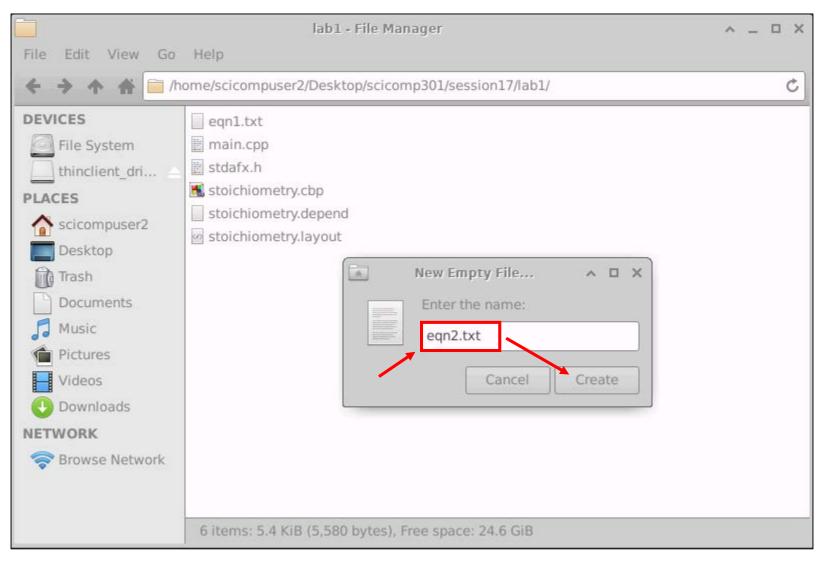
Six Columns:

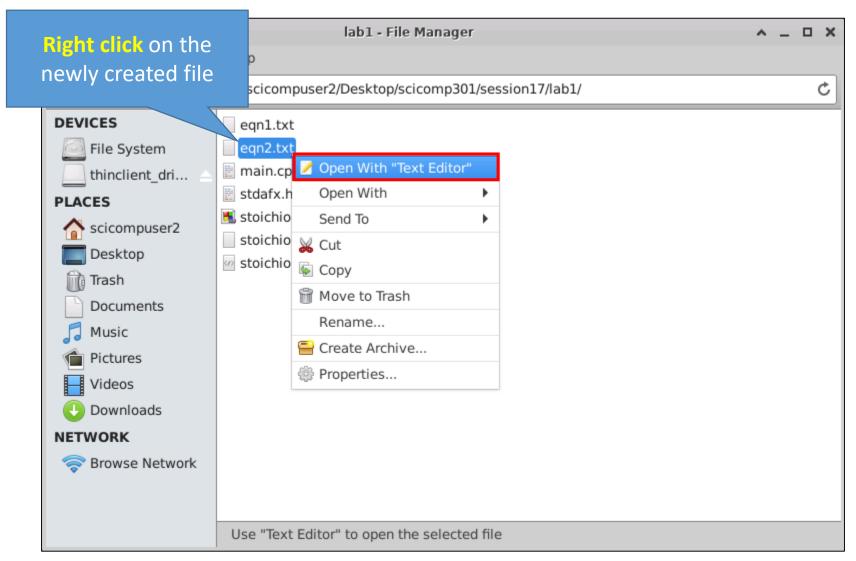
- 1. x_0
- 2. x_1
- 3. x_2
- 4. x_3
- 5. x_4
- 6. x_5

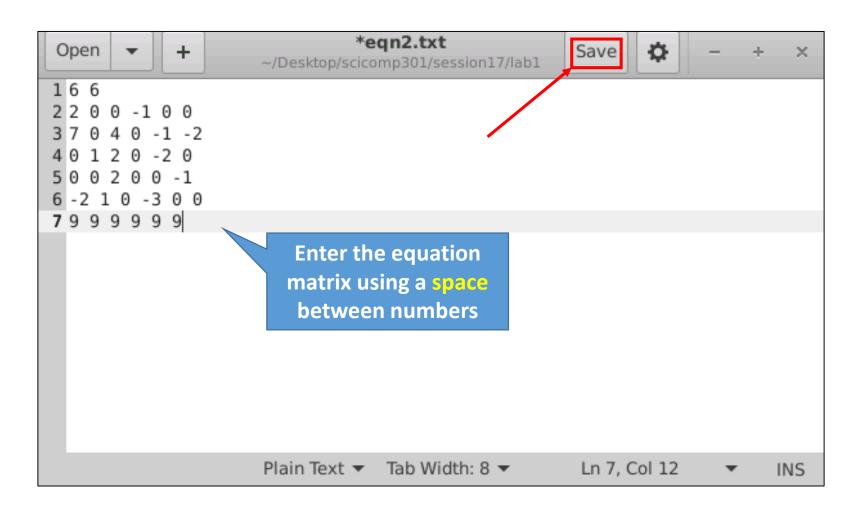
$$Cr_2O_7^{2-} + H^+ + H_2C_2O_4 \rightarrow Cr^{3+} + H_2O + CO_2$$



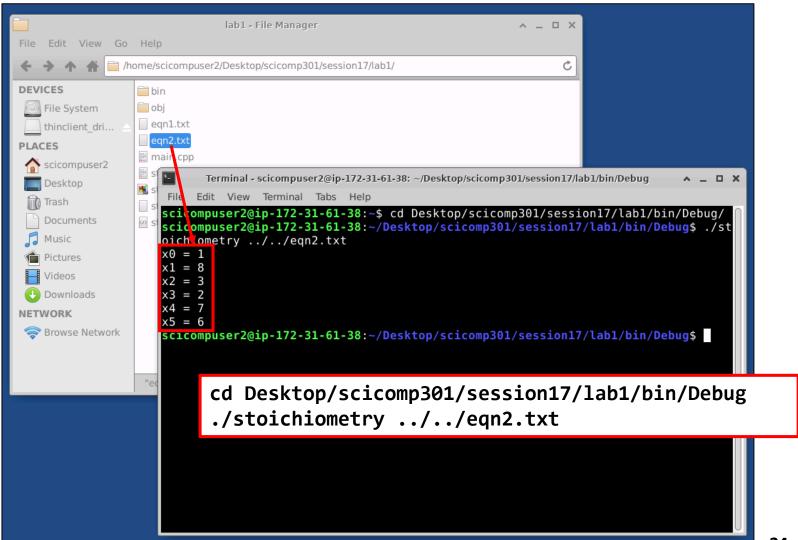






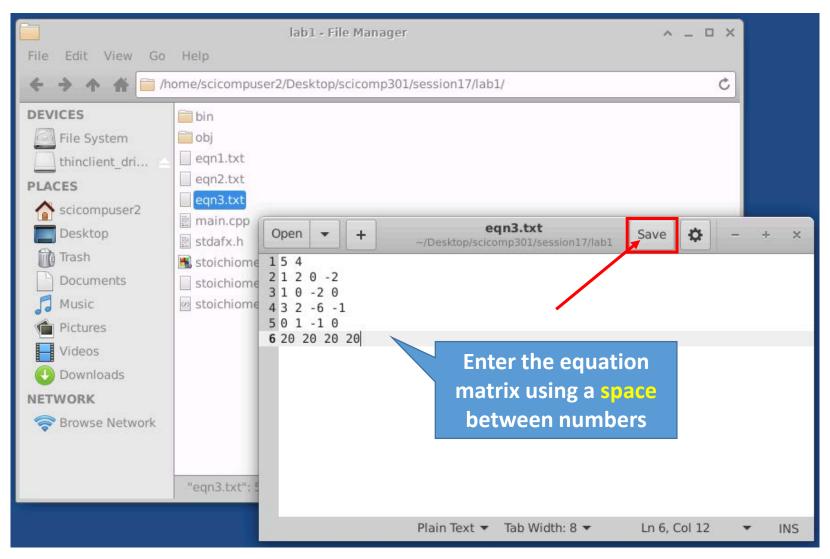


Balancing Chemical Equation 2

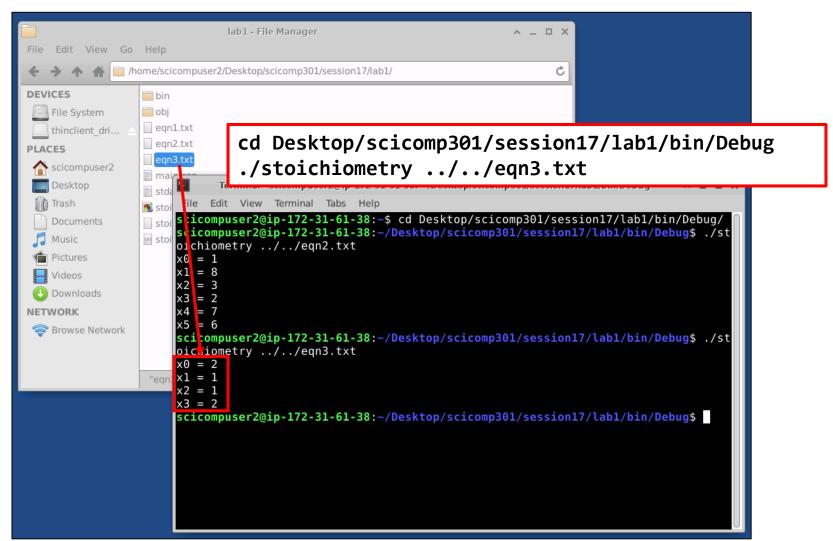


Verify Chemical Equation 2

$$\begin{array}{c} \text{HNO}_3 + \text{Ca}(\text{OH})_2 \rightarrow \text{Ca}(\text{NO}_3)_2 + \text{H}_2\text{O} \\ \hline x_0 \text{HNO}_3 + x_1 \text{Ca}(\text{OH})_2 \rightarrow x_2 \text{Ca}(\text{NO}_3)_2 + x_3 \text{H}_2\text{O} \\ \hline \text{There is no row for charges in this equation} \\ \hline \text{Hydrogen (H)} & \text{File Edit View Search Tools Document} \\ \hline \text{Hydrogen (N)} & \text{Save} \\ \hline \text{Oxygen (O)} & \text{Oxygen (O)} \\ \hline \text{$$



Balancing Chemical Equation 3



Verify Chemical Equation 3

```
x_0 \text{HNO}_3 + x_1 \text{Ca}(\text{OH})_2 \rightarrow x_2 \text{Ca}(\text{NO}_3)_2 + x_3 \text{H}_2 \text{O}_3
       Terminal - scicompusera@ip-172-31-71-74: ~/Desktop/scicomp301/session17/lab1/bin/Debug
     Edit View Terminal Tabs Help
scicompusera@ip-172-31-71-74:~$ cd_Desktop/scicomp301/session17/lab1/bin/Debug
scicompusera@ip-172-31-71-74: >/Desktop/scicomp301/session17/lab1/bin/Debug$ ./st
oichiometry ../../eqn3.txt
scicompusera@ip-172-31-71-74:~/Desktop/scicomp301/session17/lab1/bin/Debug$
            2HNO_3 + 1Ca(OH)_2 \rightarrow 1Ca(NO_3)_2 + 2H_2O
```

Avoid Writing **Brittle** Code

```
static void NestedForLoops()
   for (int x0 = 0; x0 < 50; x0++)
        for (int x1 = 0; x1 < 50; x1++)
            for (int x2 = 0; x2 < 50; x2++)
                    for (int x4 = 0; x4 < 50; x4++)
                            for (int x6 = 0; x6 < 50; x6++)
                                // Evaluate objective function
```

- We could use nested for() loops, but what if number of unknowns (x_n) changes?
- Objective function would be evaluated 50⁷ times =

• No guarantee that molar coefficients $1 \le x_n \le 50$

$$P_2I_4 + P_4 + H_2O \rightarrow PH_4I + H_3PO_4$$

vector<vector<int>>* stack;

	Index	Current	Max	
Unknown		Value	Value	
\mathbf{x}_{0}	0	1	9	
x_{1}	1	1	9	
X_2	1	1	9	< stackLevel
X_3	2	1	9	
x_4	3	1	9	
X ₅	4	1	9	

Our **stack** is a **vector of vectors**, with a row for **each unknown** and each row having **two** columns.

We also keep a pointer to the current **stack level**

Level	Range											
x ₀	1	2	3	4	5	6	7	8	9			
x1	1	2	3	4	5	6	7	8	9			
x2	1	2	3	4	5	6	7	8	9			
х3	1	2	3	4	5	6	7	8	9			
х4	1	2	3	4	5	6	7	8	9			
x5	1	2	3	4	5	6	7	8	9			

 At each iteration, the current level moves to the next position in its range, and checks the constraints

Level	Range											
x ₀	1	2	3	4	5	6	7	8	9			
x1	1	2	3	4	5	6	7	8	9			
x2	1	2	3	4	5	6	7	8	9			
х3	1	2	3	4	5	6	7	8	9			
x4	1	2	3	4	5	6	7	8	9			
x5	1	2	3	4	5	6	7	8	9			

 If the equation balances, but a new minimum atom count (sum) is found, display the current value for each unknown, and that sum becomes the new minimum

Level	Range											
x ₀	1	2	3	4	5	6	7	8	9			
x1	1	2	3	4	5	6	7	8	9			
x2	1	2	3	4	5	6	7	8	9			
х3	1	2	3	4	5	6	7	8	9			
x4	1	2	3	4	5	6	7	8	9			
x5	1	2	3	4	5	6	7	8	9			

Level	Range											
\mathbf{x}_0	1	2	3	4	5	6	7	8	9			
x1	1	2	3	4	5	6	7	8	9			
x2	1	2	3	4	5	6	7	8	9			
х3	1	2	3	4	5	6	7	8	9			
x4	1	2	3	4	5	6	7	8	9			
x5	1	2	3	4	5	6	7	8	9			

Level	Range											
x ₀	1	2	3	4	5	6	7	8	9			
x1	1	2	3	4	5	6	7	8	9			
x2	1	2	3	4	5	6	7	8	9			
х3	1	2	3	4	5	6	7	8	9			
x4	1	2	3	4	5	6	7	8	9			
x5	1	2	3	4	5	6	7	8	9			

- Just as a car odometer rolls over when a digit range reaches its limit, we increment the prior level's position and reset the current level to 1
- The program ends when every level in the stack has reached the limit of its individually assigned range

```
Find the lowest level
             bool FindSolution()
                                                                  in the stack which has
                 int stackLevel = 0;
                                                                  not yet enumerated
                 while (stackLevel >= 0) {
                      while (stack->at(stackLevel).at(0)
                                                                       its full range
                          < stack->at(stackLevel).at(1))
                          if (stackLevel == (int)stack->size() - 1) break;
 Invoke the
                          stackLevel++;
 objective
                          stack->at(stackLevel).at(0) = 1;
  function
                        (IsSolution())
                          return true;
                      stack->at(stackLevel).at(0)++;
                      while (stack->at(stackLevel).at(0)
                                                                  Walk up the stack,
Use the next
                          == stack->at(stackLevel).at(1))
                                                                 skipping all parent
value in this
                          stackLevel--;
                                                                    levels that have
                          if (stackLevel < 0) break;</pre>
level's range
                          stack->at(stackLevel).at(0)++;
                                                                  already completed
                                                                      their range
                  return false:
```

A solution is valid only if the convolution of every matrix row with every stack row == 0

```
bool IsSolution()
{
    for (size_t j{};j < eqn->size() - 1;++j) {
        int sum = 0;
        for (size_t i{};i < stack->size();++i)
            sum += eqn->at(j).at(i) * stack->at(i).at(0);

    if (sum != 0)
        return false;
    }
    return true;
}
```

Balancing Ionic Equations

- Pick <u>one</u> of the following equations to balance using the stoichiometry application
- In the Lab 1 folder, create a text file that encodes your chosen equation using the corresponding filename

eqn4.txt:
$$MnO_4^- + Fe^{2+} + H^+ \rightarrow Mn^{2+} + Fe^{3+} + H_2O$$

eqn5.txt:
$$C_8H_{18} + O_2 \rightarrow CO_2 + H_2O$$

eqn6.txt:
$$C_7H_6O_2 + O_2 \rightarrow CO_2 + H_2O$$

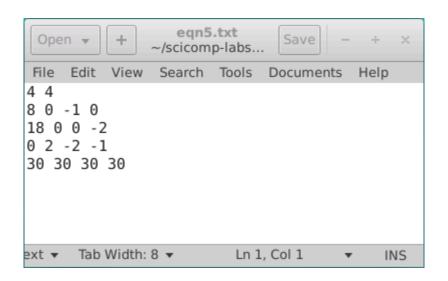
eqn7.txt:
$$P_2I_4 + P_4 + H_2O \rightarrow PH_4I + H_3PO_4$$



$$MnO_4^- + Fe^{2+} + H^+ \rightarrow Mn^{2+} + Fe^{3+} + H_2O$$

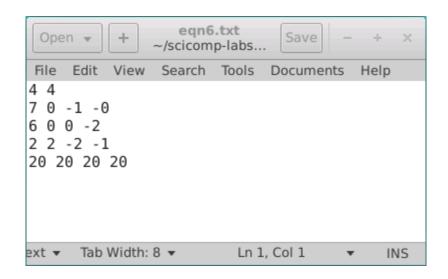
$$1MnO_4^- + 5Fe^{2+} + 8H^+ \rightarrow 1Mn^{2+} + 5Fe^{3+} + 4H_2O$$

$$C_8H_{18} + O_2 \rightarrow CO_2 + H_2O$$



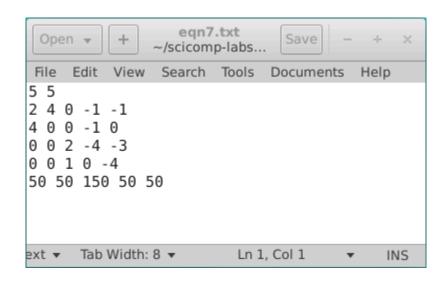
$$2C_8H_{18} + 25O_2 \rightarrow 16CO_2 + 18H_2O$$

$$C_7H_6O_2 + O_2 \rightarrow CO_2 + H_2O$$



$$2C_7H_6O_2 + 15O_2 \rightarrow 14CO_2 + 6H_2O$$

$$P_2I_4 + P_4 + H_2O \rightarrow PH_4I + H_3PO_4$$



$$10P_2I_4 + 13P_4 + 128H_2O \rightarrow 40PH_4I + 32H_3PO_4$$

Diphosphorus tetraiodide

Now you know...

https://www.webqc.org/balance.php

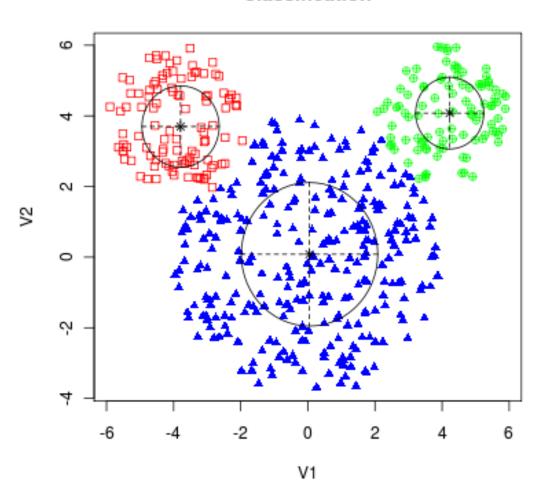
- We can encode a chemical equation into a matrix, where the rows represent the element constraints for reactants and products and the columns represent the unknown coefficients of the molecules
- We can transform balancing ionic equations to a linear program, which can be solved using brute-force if the search space is sufficiently small
- Even relatively simple equations can exponentially increase the search space & using nested for() loops makes the code too brittle
- It is better to use the **Simplex** method, and normalize any rational solutions back to integers for the final answer

Session Goals

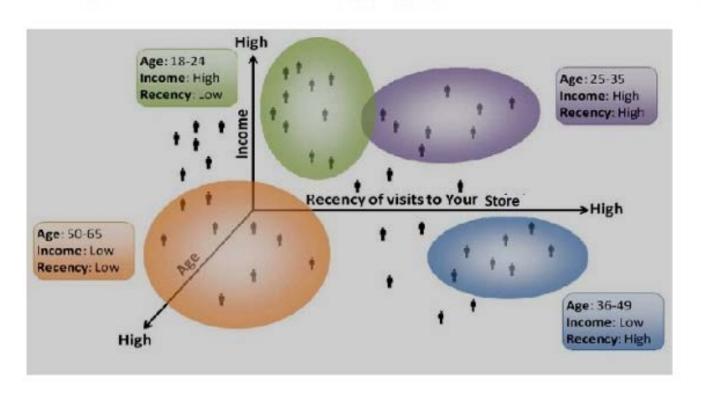
- Study an efficient algorithm for grouping similar data sets
- Gain an appreciation for k-means Clustering an algorithm that iterates by assigning data points to a cluster
 - The goal is to programmatically assign data points that are "near" each other to the same cluster so we can perform group statistics on data that "belongs" together
 - Learn the limitations of k-means and visualize the dangers of "over fitting" data
 - Envision how we can use a **mean** (μ) **distance cut-off** to programmatically identify **data outliers**

- k-means clustering aims to partition n observations into k clusters (where each cluster has its own μ point)
- ullet Each observation is assigned to the cluster whose μ point is closest to that observation's point
- The problem is computationally difficult (NP-hard); however, there are efficient heuristic algorithms that are commonly employed and converge quickly to a local optimum
- The term "k-means" was first used by James MacQueen in 1967, though the idea goes back to Hugo Steinhaus in 1957
- A more efficient version was published Hartigan and Wong in 1979

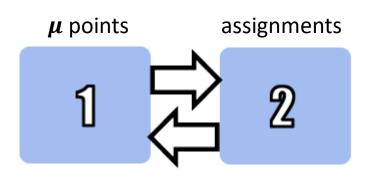
Classification

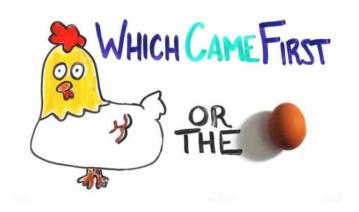


Example - Clusters using Age, Income & Recency



- A one-to-many mapping between clusters and data points
 - A cluster is a set of one or more unique data points
 - Every data point must be <u>assigned</u> to only one cluster at a time
 - Data points are initially assigned to clusters in a round-robin fashion
- The μ point of each cluster is its geometric center (centroid)
 calculated by taking the average value (in each dimension) of
 only the data points assigned to that cluster
- Assignments can change the μ points... which then can change the assignments... which then can change the μ points... which then can change the assignments... which then can...





$$G_{\mu\nu} + \Lambda T_{\mu\nu} = \frac{8\pi G}{c^4} T_{\mu\nu}$$

Spacetime tells matter how to move; matter tells spacetime how to curve.

— John Archibald Wheeler —

- The algorithm uses **two** phases per round:
 - 1. For each cluster, a new centroid (μ point) is calculated using the <u>current</u> data point assignments
 - If the new $\mu \neq$ current μ then move cluster so current $\mu =$ new μ
 - 2. For each data point, find the cluster which has the closest μ point using Pythagorean distance formula
 - If necessary, reassign the data point to the new closest cluster subject to constraint that every cluster must have one data point
- The algorithm repeats phases #1 and #2 until no cluster moves and no data point is reassigned
- When the algorithm reaches steady-state the cluster has "converged"

Open Lab 2 – kMeansClustering.h

```
class DataPoint
public:
    DataPoint() = default;
    DataPoint(double new x, double new y);
    double x{};
    double v{}:
    Cluster* c{nullptr};
                                        class Cluster
              Every datapoint
                                        public:
            belongs to just one
                                            Cluster(int index);
                                            double x{};
                  cluster
                                            double y{};
                                            string clr{};
                                             int population{};
                                            double mean distance{};
```

View Lab 2 – kMeansClustering.h

```
class DataPoint
{
public:
    DataPoint() = default;
    DataPoint(double new_x, double new_y);
    double x{};
    double y{};
    Cluste* c{nullptr};
};
```

The μ point (center) of a cluster is the average of the coordinates (x, y) of all the datapoints that belong to that cluster

Every datapoint has an (x, y) Cartesian coordinate

View Lab 2 – kMeansClustering.h

```
class DataPoint
{
public:
    DataPoint() = default;
    DataPoint(double new_x, double new_y);
    double x{};
    double y{};
    Cluster* c{nullptr};
};
```

```
class Cluster
{
public:
    Cluster(int index);
    double x{};
    double y{};
    string clr{};
    int population{};
    double mean_distance{};
};
```

View Lab 2 – kMeansClustering.h

```
class DataPoint
{
public:
    DataPoint() = default;
    DataPoint(double new_x, double new_y);
    double x{};
    double y{};
    Cluster* c{nullptr};
};
```

```
class Cluster
{
public:
    Cluster(int index);
    double x{};
    double y{};
    string clr{};
    int population{};
    double mean_distance{};
};
```

Open Lab 2 – main.cpp

```
main.cpp 🗷 kMeansClustering.cpp
                                      kMeansClustering.h
                                                           X
          #include "stdafx.h"
          #include "simplescreen.h"
          #include "kMeansClustering.h"
                                                   Global variables are
                                                   declared outside the
          using namespace std;
                                                  scope of any function
         int num clusters{3}
          double mean multiple{0};
  10
          vector<DataPoint*>* dataPoints{};
         vector<Cluster*>* clusters{};
  11
  12
         bool converged{};
  13
          double GetDistance(double x1, double y1, double x2, double y2)
  14
  15
             return sqrt(pow(x1-x2,2) + pow(y1-y2,2))
  16
  17
```

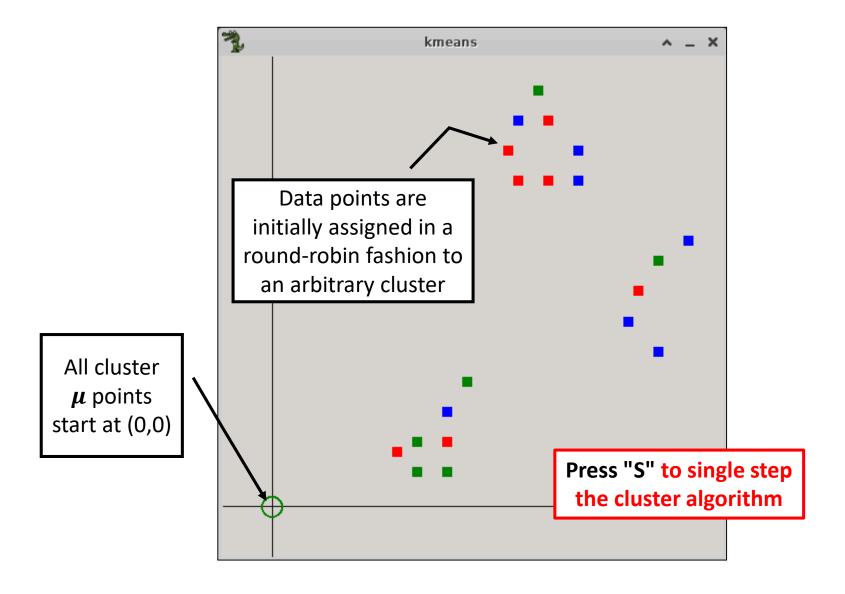
Even k-means uses the Pythagorean theorem!

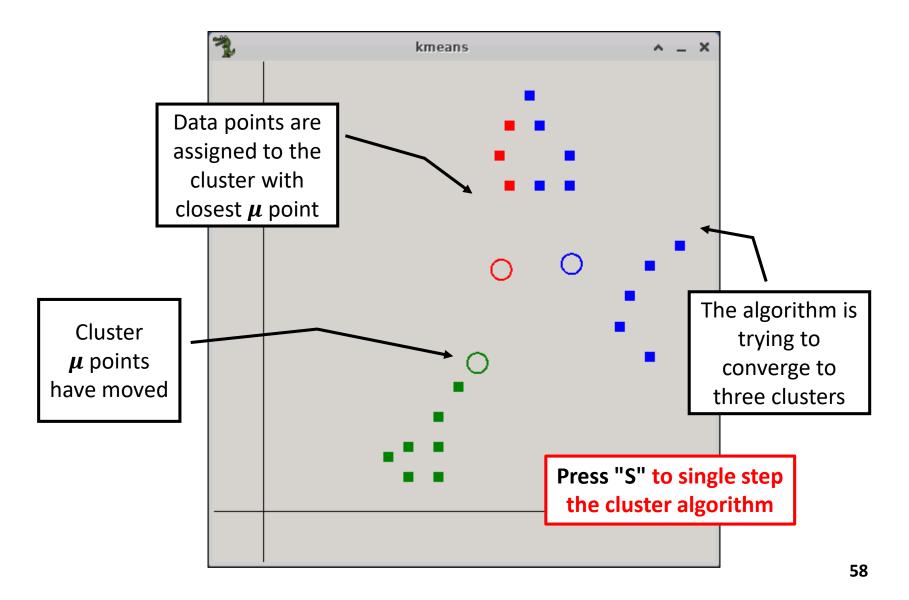
```
main.cpp 🗷 kMeansClustering.cpp
                                                        kMeansClustering.h
                                                                            ×
                    13
                    14
                           double GetDistance(double x1, double y1, double x2,
                    15
                         □{
                    16
                               return sqrt(pow(x1-x2,2) + pow(y1-y2,2));
                    17
                    18
                                                                                           (x,y)
                    19
                           void InitDataPoints()
                    20
                                                                                       coordinates
                    21
                               dataPoints = new vector<DataPoint*>();
                    22
                               dataPoints->push back(new DataPoint(23, 35));
                                                                                       of each data
                               dataPoints->push back(new DataPoint(35, 18));
                    23
                                                                                           point
                    24
                               dataPoints->push back(new DataPoint(14, 3));
                    25
                               dataPoints->push back(new DataPoint(17, 6));
                    26
                               dataPoints->push back(new DataPoint(38, 15));
                    27
                               dataPoints->push back(new DataPoint(26, 41));
                    28
                               dataPoints->push_back(new DataPoint(27, 38));
                    29
                               dataPoints->push back(new DataPoint(30, 35));
                    30
                               dataPoints->push back(new DataPoint(19, 12));
                    31
                               dataPoints->push back(new DataPoint(24, 32));
                    32
                               dataPoints->push back(new DataPoint(41, 26));
                               dataPoints->push back(new DataPoint(38, 24));
                               dataPoints->push back(new DataPoint(36, 21));
This point is a data
                               dataPoints->push back(new DataPoint(30, 32));
 outlier which we
                               dataPoints->push back(new DataPoint(17, 3));
                               dataPoints->push back(new DataPoint(12, 5));
   will use later
                               dataPoints->push back(new DataPoint(24, 38));
                               dataPoints->push back(new DataPoint(14, 6));
                               dataPoints->push_back(new DataPoint(27, 32));
                               dataPoints->push back(new DataPoint(17, 9));
                    41
                    42
                               //dataPoints->push back(new DataPoint(5, 40))
                    43
```

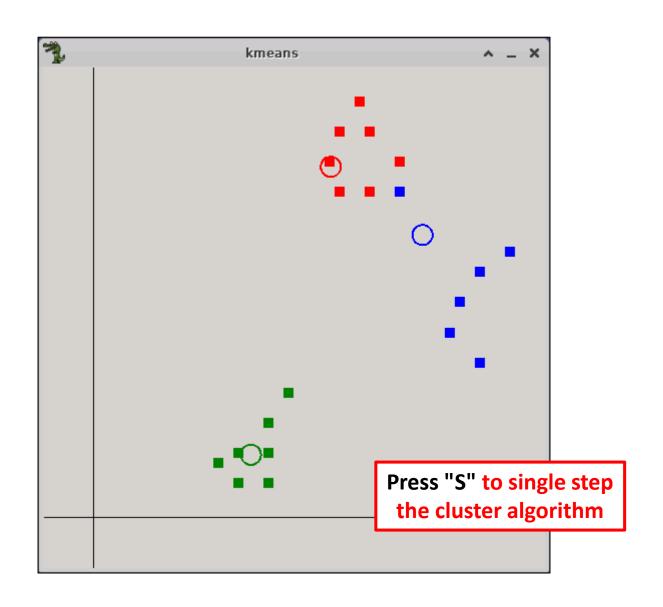
```
void InitClusters()
    clusters = new vector<Cluster*>();
    for (int i{}; i < num clusters; ++i)</pre>
        clusters->push back(new Cluster(i));
    // Assign each data point to an initial
    // cluster in a round-robin fashion
    for (size t i{}; i < dataPoints->size(); ++i)
        Cluster* c = clusters->at(i % clusters->size());
        dataPoints -> at(i) -> c = c;
        c->population++;
                                        Datapoints are assigned
                                         to an initial cluster in a
                                          round-robin fashion
```

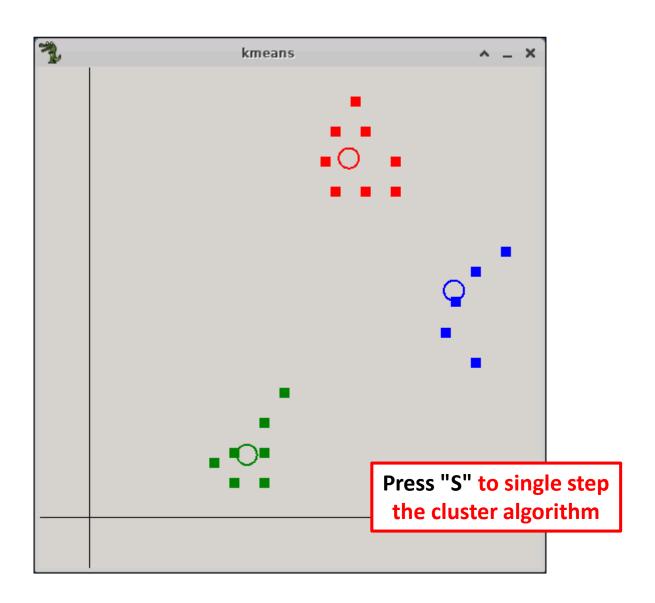
```
void draw(SimpleScreen& ss)
                                           ss.Clear():
int main()
                                           ss.DrawAxes();
                                           // Draw each data point
    SimpleScreen ss(draw, eventHandler
                                           for (auto dp : *dataPoints)
    ss.SetWorldRect(-5, -5, 45, 45);
                                               ss.DrawRectangle(dp->c->clr, dp->x, dp->y, 1, 1, 0, true);
                                           // Draw each cluster's geometric center (the "mean")
    cout << "Press S to single step th
                                           for (auto c : *clusters)
    cout << "Press Q to quit the appli
                                               ss.DrawCircle(c->x, c->y, 1, c->clr, 2);
                                           ss.Update();
    InitDataPoints();
    InitClusters();
                            void eventHandler(SimpleScreen& ss, ALLEGRO EVENT& ev)
   ss.HandleEvents()
                                if (ev.type == ALLEGRO EVENT KEY CHAR)
    return 0;
                                    switch (ev.keyboard.keycode)
                                    case ALLEGRO KEY S:
                                         UpdateClusters();
                                         ss.Redraw();
                                         if (converged)
                                             cout << "Cluster has converged" << endl;</pre>
                                         break:
                                     case ALLEGRO KEY Q:
                                         ss.Exit();
                                         break;
```

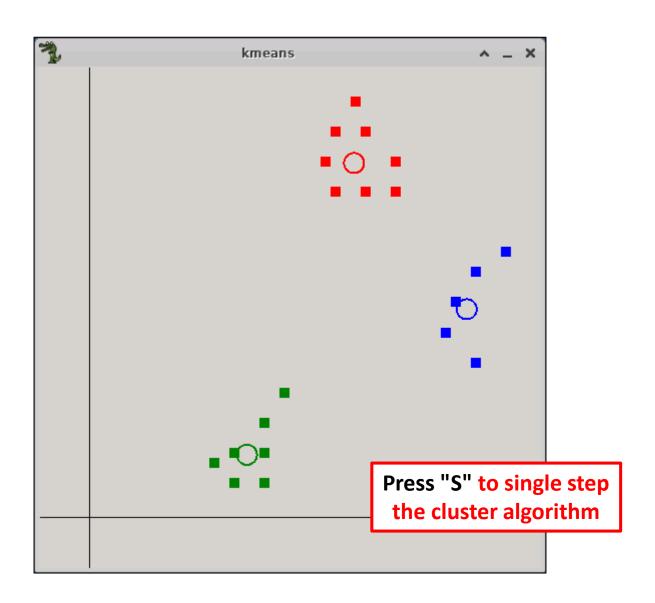
Run Lab 2 - Round 0

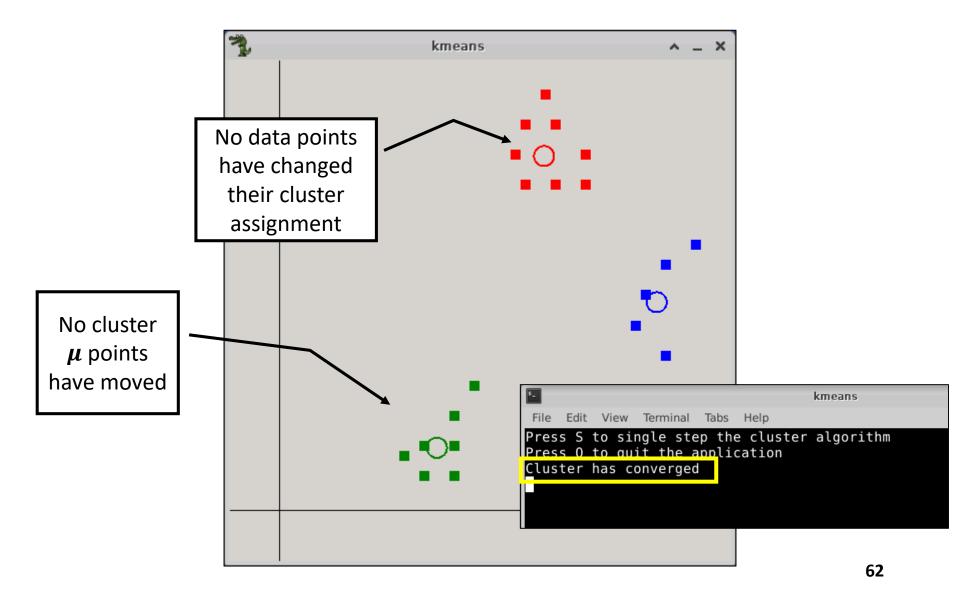




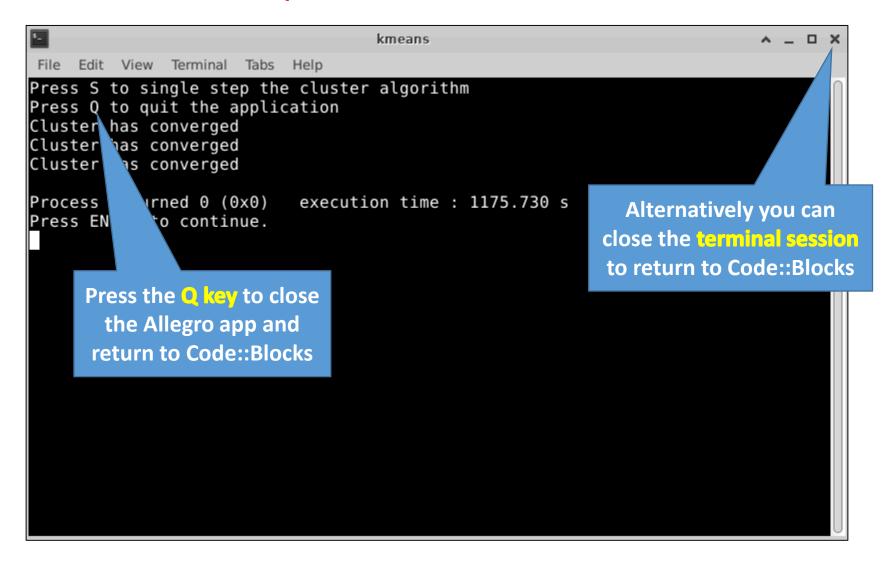


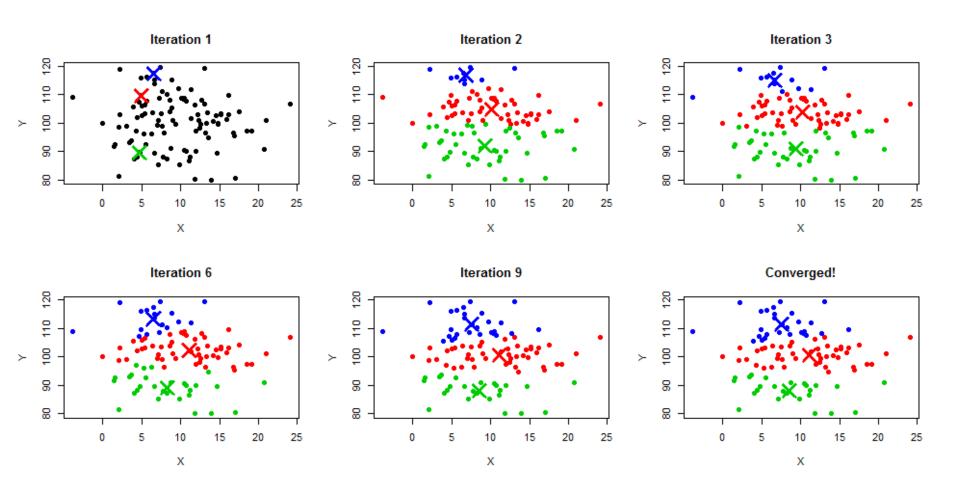




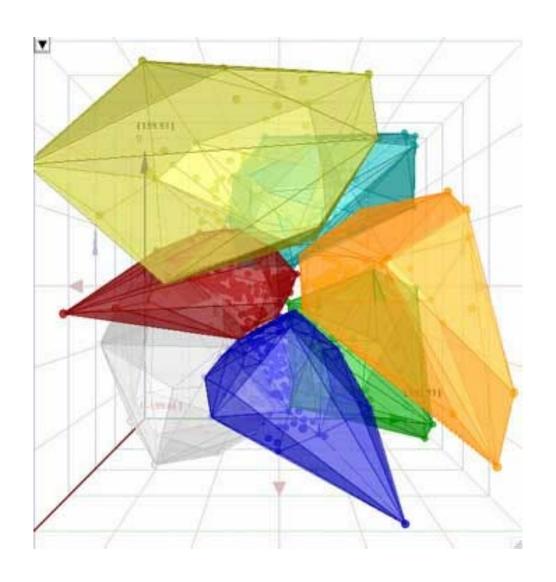


Quit Lab 2 - Round 5





n-Dimensional k-means Clustering



```
void UpdateClusters()
    // Phase I: Calculate the new geometric mean of each
    // cluster based upon current data point assignments
    bool phase1 change{false};
    for (auto c : *clusters)
        DataPoint newCenter:
        double count{}:
        for (auto dp : *dataPoints)
            if (dp->c == c)
                newCenter.x += dp->x;
                newCenter.y += dp->y;
                count++;
        newCenter.x /= count:
        newCenter.v /= count:
        // Move cluster center (mean) if necessary
       if (newCenter.x != c->x || newCenter.y != c->y)
            c->x = newCenter.x:
            c->y = newCenter.y;
            phase1 change = true;
```

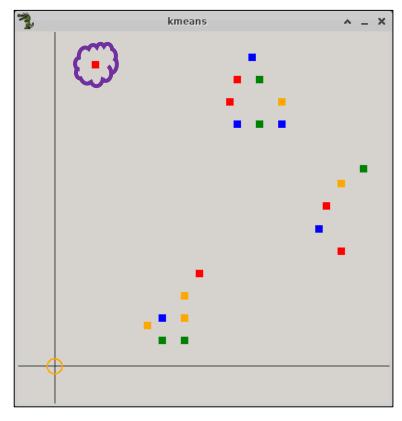
```
// Phase II: Assign data points to nearest cluster
bool phase2 change{false};
for (auto dp : *dataPoints)
    double distMin = numeric limits<double>::max();
   Cluster* cNearest = nullptr;
    for (auto c : *clusters)
        double dist = GetDistance(dp->x, dp->y, c->x, c->y);
        if (dist < distMin)</pre>
            distMin = dist:
            cNearest = c:
    // If the nearest cluster to this point is NOT
   // the point's currently assigned cluster...
   if (cNearest != dp->c)
        // Then reassign data point to the new cluster, but only
       // if this is not the last point the current cluster.
        // Reassignment must never result in an empty cluster
       if (dp->c->population > 1)
            dp->c->population--;
            // Assign point to new (nearer) cluster
            dp->c = cNearest;
            dp->c->population++;
            phase2 change = true;
if (mean multiple == 0 && !phase1 change && !phase2 change && !converged)
    converged = true;
```

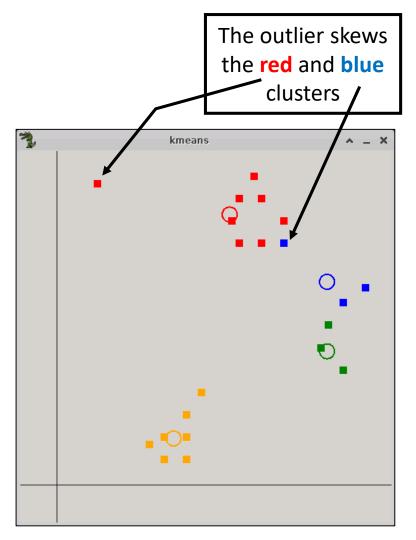
Edit Lab 2 – main.cpp

```
main.cpp 🗷
         #include "stdafx.h"
         #include "simplescreen.h"
                                         main.cpp 🗷
         #include "kMeansClustering.h"
                                                   void InitDataPoints()
                                            19
                                            20
         using namespace std;
                                            21
                                                       dataPoints = new vector<DataPoint*>();
                                            22
                                                       dataPoints->push back(new DataPoint(23, 35));
         int num clusters{4};
                                            23
                                                       dataPoints->push back(new DataPoint(35, 18));
         double mean mult/ple{0};
                                                       dataPoints->push_back(new DataPoint(14, 3));
                                            24
                                            25
                                                       dataPoints->push back(new DataPoint(17, 6));
  10
         vector<DataPoi
                          *>* dataPoints
                                            26
                                                       dataPoints->push back(new DataPoint(38, 15));
         vector<Cluste
  11
                           clusters{}:
                                            27
                                                       dataPoints->push back(new DataPoint(26, 41));
         bool conver
  12
                                                       dataPoints->push back(new DataPoint(27, 38));
                                            28
                                            29
                                                       dataPoints->push back(new DataPoint(30, 35));
                                            30
                                                       dataPoints->push back(new DataPoint(19, 12));
           Increase
                                                       dataPoints->push back(new DataPoint(24, 32));
                                            31
                                            32
                                                       dataPoints->push back(new DataPoint(41, 26));
     num clusters to 4
                                            33
                                                       dataPoints->push back(new DataPoint(38, 24));
                                            34
                                                       dataPoints->push_back(new DataPoint(36, 21));
                                            35
                                                       dataPoints->push back(new DataPoint(30, 32));
                                                       dataPoints->push back(new DataPoint(17, 3));
                                            36
                                                       dataPoints->push back(new DataPoint(12, 5));
                                            37
                                                       dataPoints->push back(new DataPoint(24, 38));
                                            38
                                                       dataPoints->push back(new DataPoint(14, 6));
                                            39
                                            40
                                                       dataPoints->push back(new DataPoint(27, 32));
 Uncomment line 42 to
                                                       dataPoints->nush_back(new_DataPoint(17, 9)):
  include a data outlier
                                                       dataPoints->push back(new DataPoint(5, 40));
```

Run Lab 2 – Data Outlier

Press "S" to single step the cluster algorithm



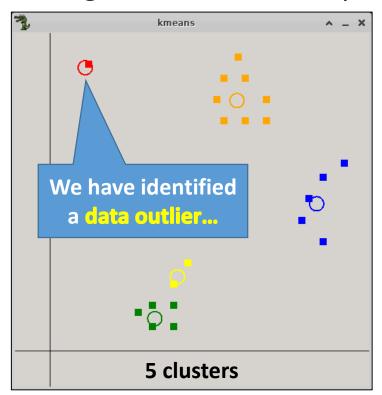


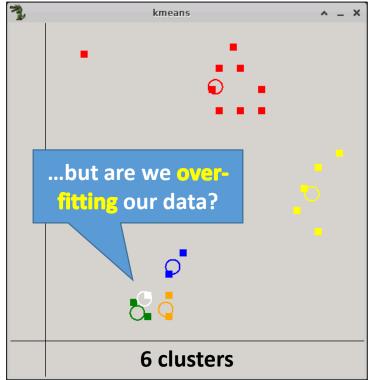
Round 0

Round 3

Edit Lab 2 – Research Questions

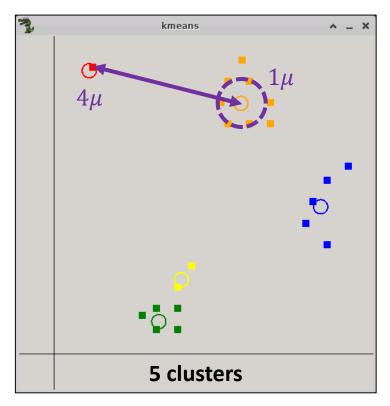
- Compare convergence with num_clusters set to 5 versus 6
- Is using more clusters a key to identifying data outliers?

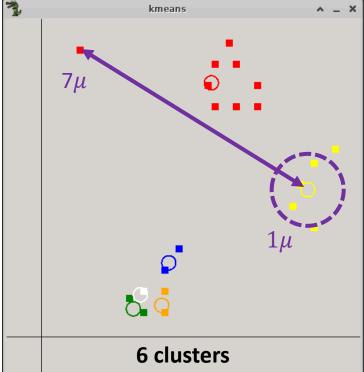




Identifying Data Outliers

• Could a mean (μ) distance cut-off be used to gauge if a datapoint reasonably belongs to a given cluster?

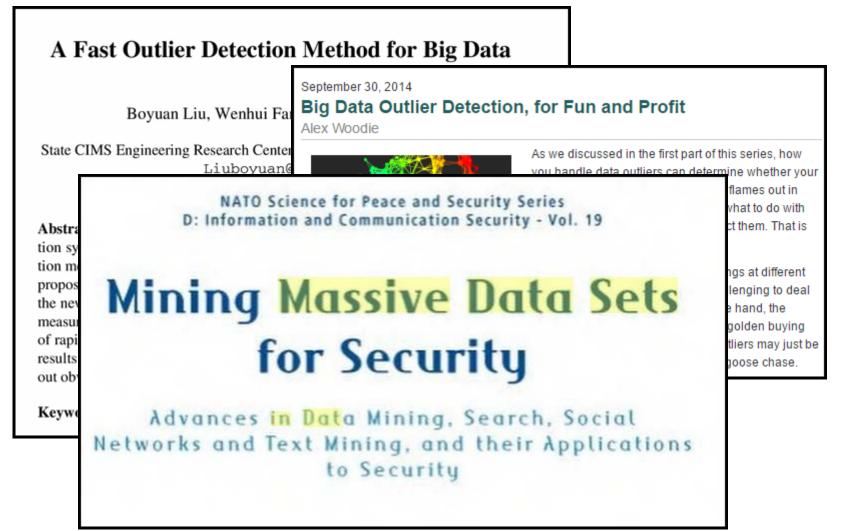




Identifying Data Outliers

- Once the algorithm converges, possible data outliers can be identified by evicting any points that are too far away from the center of their assigned cluster
 - Reasonable data points should fall around the cluster's center point in a **normal** distribution. This means that **99.97**% of data points should be less than 3σ away from the center of their cluster
 - Upon convergence, if the distance between a data point and the center of its cluster is greater than some **multiplier** of that cluster's mean (μ) spread, it probably is an outlier and should be evicted
 - If any data point is evicted, the algorithm must continue to iterate
- If a data point ultimately gets evicted from all clusters, it is probably a data capture error or a statistical anomaly
- This technique will be explored in Exam #3

Identifying Data Outliers



Now you know...

- We can use k-means clustering to categorize data points by its nearness to a similar group's statistical means
 - Data sets which are homogenous (have a more uniform distribution of values) may not converge to a meaningful set of clusters since the data lacks sharpness
 - We can use k-means to identify data outliers so we can avoid allowing those anomalies from skewing the main observations – this process is called evicting a point from the cluster
- We can extend k-means clustering to analyze multidimensional data sets to look for trends
 - This is exactly how Amazon and Netflix provide you purchase recommendations based upon your shared interests & tastes compared with other buyers – a very effective selling technique!