```
In [47]: # 1
              using LinearAlgebra
              using Statistics
              import Random
              using Plots
              using VMLS
              Random.seed!(8)
              X = vcat([[0, -0.2] + 0.1*randn(2) for i = 1:30],
                [ [0.5, 0.5] + 0.1*randn(2)  for i = 1:30 ],
                [ [0.5, -0.5] + 0.1*randn(2)  for i = 1:20 ],
                [ [0, 0.4] + 0.06* randn(2)  for i = 1:20])
              Che_res = [0.2, 0]
              scatter([x[1] for x in X], [x[2] for x in X])
              scatter!([Che_res[1]], [Che_res[2]])
              plot!(legend = false, grid = false, size = (500,500))
Out[47]:
                  0.6
                  0.4
                  0.2
                  0.0
                 -0.2
                 -0.4
                 -0.6
                     -0.2
                                       0.0
                                                         0.2
                                                                                            0.6
In [48]: # 2 a, b
              distances_1 = zeros(length(X))
              for i in 1:length(X)
                    distances_1[i] = norm(X[i] - Che_res)
              println("These are all distances (in terms of 2-norm) between Chelsea's restaurant and all her customers: \n\n", dist
              println("\nCustomer ", findmin(distances_1)[2], " is the closest customer to Chelsea's restaurant in terms of 2-nor
              These are all distances (in terms of 2-norm) between Chelsea's restaurant and all her customers:
              [0.2009677473503477,\ 0.25896260382301883,\ 0.42061324214195295,\ 0.25681145113350495,\ 0.18311296675364602,\ 0.3337325170]
              375205, 0.30235710582270836, 0.34911442689820105, 0.42816530554960197, 0.22796247114875975, 0.2735099500161391, 0.408
              7802805445024, 0.3040375580970352, 0.33414103904197495, 0.18072514224273636, 0.25515907137194827, 0.2968302097625538,
              0.45286928460732195,\ 0.3234524421184,\ 0.38190313311155316,\ 0.35804600583071267,\ 0.2214777808533561,\ 0.271509228858504
              15, 0.37665689382076706, 0.41468352729322994, 0.24075740733699408, 0.30905153011315495, 0.21752699574550757, 0.207047
              14569444144, \ 0.10825853896279175, \ 0.444915768740275, \ 0.6106488410454415, \ 0.5312820897959437, \ 0.6122261689301772, \ 0.59812820897959437, \ 0.6122261689301772, \ 0.59812820897959437, \ 0.6122261689301772, \ 0.59812820897959437, \ 0.6122261689301772, \ 0.59812820897959437, \ 0.6122261689301772, \ 0.59812820897959437, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.6122261689301772, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.612226168930172, \ 0.61222616893017
              79904266677406, 0.5766584137636611, 0.35393194713041715, 0.6718186266177286, 0.46656702131250816, 0.5196961812413883,
              0.6136070015320558,\ 0.6759016621073926,\ 0.49828951069410593,\ 0.7497210949018761,\ 0.659654323821737,\ 0.465923012053291
              8, 0.5244285277873351, 0.6781799833146084, 0.3714505462254292, 0.7015698508679072, 0.43039281510751, 0.73811138454117
              1, \ 0.7736371010035272, \ 0.5157608183668018, \ 0.7636073781063303, \ 0.7327114255881644, \ 0.5223263796086925, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970772746, \ 0.595970774746, \ 0.595970772746, \ 0.5959707744, \ 0.5959707744, \ 0.59597
              6451,\ 0.5472016474256196,\ 0.5418461028791919,\ 0.7230630997432982,\ 0.6776574061779677,\ 0.5082080205663847,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.6364061470,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\ 0.636406147,\
              2142769535, 0.7170718126318616, 0.525251441579659, 0.6733385934672023, 0.5552178155141244, 0.6177618325779338, 0.6874
              465476652216085, 0.45597130863713103, 0.5182181548606503, 0.3833352254578807, 0.5066593191106892, 0.3672579758821926
              7, 0.5585006119512392, 0.40777565446743036, 0.6161698869830765, 0.5131595037969192, 0.4111397415266218, 0.46217438505
              98974, 0.39548945948269726, 0.4774786944055352, 0.5029716239002253, 0.5933484517569716, 0.4392977378986075, 0.4680315
              189077585, 0.5119191981469117]
              Customer 30 is the closest customer to Chelsea's restaurant in terms of 2-norm.
In [49]: # 2 c
              distances_2 = zeros(length(X))
              for i in 1:length(X)
                    distances_2[i] = abs(X[i][1] - Che_res[1]) + abs(X[i][2] - Che_res[2])
              # println(distances_2)
              println("Customer ", findmax(distances_2)[2], " is the farthest customer to Chelsea's restaurant in terms of 1-nor
              Customer 65 is the farthest customer to Chelsea's restaurant in terms of 1-norm.
In [50]: # 2 d 2-norm
              distances_1 = zeros(length(X))
              for i in 1:length(X)
                    distances_1[i] = norm(X[i] - Che_res)
              copy_distances_1 = copy(distances_1)
              small_index_1 = []
              for i in 1:5
                    fake_index = findmin(copy_distances_1)[2]
                    real_index = findall(x -> x==copy_distances_1[fake_index], distances_1)
                    append!(small_index_1, real_index)
                    deleteat!(copy_distances_1, findmin(copy_distances_1)[2])
              println(small_index_1)
              Any[30, 15, 5, 1, 29]
In [51]: # 2 d 1-norm
              distances_2 = zeros(length(X))
              for i in 1:length(X)
                     distances_2[i] = abs(X[i][1] - Che_res[1]) + abs(X[i][2] - Che_res[2])
              copy_distances_2 = copy(distances_2)
              small_index_2 = []
              for i in 1:5
                     fake_index = findmin(copy_distances_2)[2]
                    real_index = findall(x -> x==copy_distances_2[fake_index], distances_2)
                    append!(small_index_2, real_index)
                    deleteat!(copy_distances_2, findmin(copy_distances_2)[2])
              println(small_index_2)
              Any[30, 1, 15, 5, 16]
              Answer 2d.: The 5 closest customers to Chelsea's restaurant calculated using 1-norm and 2-norm are "not" the
In [52]: # 3
              Random.seed!(8)
              y = randn(100)
              Random.seed!(8)
              quality = y * 0.5 + rand(100) * 0.1
              Random.seed!(10)
              nutrition = y * 0.1 + rand(100) * 0.2
              price = quality * 1 + nutrition * 0.1 + rand(100) * 2
              predicted_price = 0.5*quality + 7.1*nutrition .+ 0.1
              println("Data point ", findmax([norm(price[i] - predicted_price[i]) for i = 1:length(price)])[2], " is the farthest
                (in terms of 2-norm) to its prediction.")
              scatter(price, predicted_price)
              scatter!([price[97]], [predicted_price[97]])
              plot!(legend = false, grid = false, size = (500,500), xaxis = "True Price", yaxis = "Predicted Price")
              Data point 97 is the farthest (in terms of 2-norm) to its prediction.
Out[52]:
                Predicted Price
                                           0
                        -1
                                                         True Price
In [53]: # 4 k-means algorithm
              function kmeansfunction(x, k; maxiters = 100, tol = 1e-5)
              N = length(x) # number of points to be cluster
              n = length(x[1]) # the dimension
              distances = zeros(N) # used to store the distance of each point to the nearest representative.
              reps = [zeros(n) for i=1:k] # used to store representatives.
              # 'assignment' is an array of N integers between 1 and k.
               # The initial assignment is chosen randomly.
              assignment = [rand(1:k) for i=1:N]
              Jprevious = Inf # used in stopping condition
              for iter = 1:maxiters
                    # Cluster j representative is average of points in cluster j.
                    for j = 1:k
                          group = [i for i=1:N if assignment[i]==j] # get which point index belongs to the group j
                          reps[j] = sum(x[group]) / length(group); # reps is the clustering center
                    # For each x[i], find distance to the nearest representative
                    # and its group index.
                    for i = 1:N
                           (distances[i], assignment[i]) = findmin([norm(x[i] - reps[j]) for j=1:k])
                    end;
                    # Compute clustering objective.
                    J = norm(distances)^2 / N # distance to the clustering center
                    # Show progress and terminate if J stopped decreasing.
                    println("Iteration ", iter, ": Jclust = ", J, ".")
                    if (iter > 1) && (abs(J - Jprevious) < tol*J)</pre>
                            return assignment, reps
                    end
                Jprevious = J
                end
                end
Out[53]: kmeansfunction (generic function with 1 method)
In [54]: # 4 a, b
              assignment, reps = kmeansfunction(X, 3)
              println("\n\n", reps, " are the best locations to open these three restaurants for Chelsea's franchisee.")
              Iteration 1: Jclust = 0.21805080743372227.
              Iteration 2: Jclust = 0.054753907704250544.
              Iteration 3: Jclust = 0.04740440133730133.
              Iteration 4: Jclust = 0.04707074206229065.
              Iteration 5: Jclust = 0.04707074206229065.
              [[0.5371909960116741, -0.5161171797997566], [0.02958765596728597, -0.21210401348889638], [0.2808290177102955, 0.46611]
              675910208605]] are the best locations to open these three restaurants for Chelsea's franchisee.
In [55]: k = 3 # Number of groups
              N = length(X) # Total number of points to be clustered
              grps = [[X[i] for i=1:N if assignment[i] == j] for j=1:k]
              scatter([c[1] for c in grps[1]], [c[2] for c in grps[1]])
              scatter!([c[1] for c in grps[2]], [c[2] for c in grps[2]])
              scatter!([c[1] for c in grps[3]], [c[2] for c in grps[3]])
              scatter!([c[1] for c in reps], [c[2] for c in reps])
              plot!(legend = false, grid = false, size = (500,500))
Out[55]:
                  0.6
                  0.4
                  0.2
                  0.0
                 -0.4
                 -0.6
                                       0.0
                                                         0.2
                     -0.2
In [56]: # 4 c
              # 4 k-means algorithm with 1-norm
              function kmeansfunction_2(x, k; maxiters = 100, tol = 1e-5)
              N = length(x) # number of points to be cluster
              n = length(x[1]) # the dimension
              distances = zeros(N) # used to store the distance of each point to the nearest representative.
              reps = [zeros(n) for i=1:k] # used to store representatives.
              # 'assignment' is an array of N integers between 1 and k.
              # The initial assignment is chosen randomly.
              assignment = [rand(1:k) for i=1:N]
              Jprevious = Inf # used in stopping condition
              for iter = 1:maxiters
                    # Cluster j representative is average of points in cluster j.
                    for j = 1:k
                           group = [i for i=1:N if assignment[i]==j] # get which point index belongs to the group j
                           reps[j] = sum(x[group]) / length(group); # reps is the clustering center
                    # For each x[i], find distance to the nearest representative
                    # and its group index.
                    for i = 1:N
                           (distances[i], assignment[i]) = findmin([(abs(x[i][1] - reps[j][1]) + abs(x[i][2] - reps[j][2]))  for j=1:k])
                    end;
                    # Compute clustering objective.
                    J = norm(distances)^2 / N # distance to the clustering center
                    # Show progress and terminate if J stopped decreasing.
                    println("Iteration ", iter, ": Jclust = ", J, ".")
                    if (iter > 1) && (abs(J - Jprevious) < tol*J)</pre>
                            return assignment, reps
                    end
                Jprevious = J
Out[56]: kmeansfunction_2 (generic function with 1 method)
In [57]: assignment, reps = kmeansfunction_2(X, 3)
              Iteration 1: Jclust = 0.3403702703823343.
              Iteration 2: Jclust = 0.09979403232802793.
              Iteration 3: Jclust = 0.09823290083113136.
              Iteration 4: Jclust = 0.09788499721599553.
              Iteration 5: Jclust = 0.09773526641653874.
              Iteration 6: Jclust = 0.09773526641653874.
Out[57]: ([1, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, ... 1, 1, 1, 1, 1, 1, 1, 1, 1], [[-0.02373170768873549, 0.33606929730232826], [0.25]
              79074365705538, -0.35895024685859395], [0.4792906462350338, 0.5022140724893905]])
In [58]: k = 3 # Number of groups
              N = length(X) # Total number of points to be clustered
              grps = [[X[i] for i=1:N if assignment[i] == j] for j=1:k]
              scatter([c[1] for c in grps[1]], [c[2] for c in grps[1]])
              scatter!([c[1] for c in grps[2]], [c[2] for c in grps[2]])
              scatter!([c[1] for c in grps[3]], [c[2] for c in grps[3]])
              plot!(legend = false, grid = false, size = (500,500))
Out[58]:
                  0.6
                  0.2
                  0.0
                 -0.4
                 -0.6
                     -0.2
                                       0.0
                                                         0.2
                                                                          0.4
                                                                                            0.6
              Answer 4c. :
              The plot using 1-norm to show the customers' locations on the map is different from (b).
              The plot in (b) seperates the customers into "upper group", "lower left group", and "lower right group".
              However, the plot in (c) seperates the customers into "bottom group", "upper left group", and "upper right
              group".
              Answer 4d. :
              3 is not a good clustering number because apparently there are "4" seperated groups on the map.
              4 would be a good number.
In [59]: # 5
              function Gram_Schmidt(a; tol = 1e-10)
                    copy_a = copy(a)
                    copy_a = convert(Vector{Vector{Float64}}, copy_a)
                    for i = 1:length(copy_a)
                           for j = 1:i-1
                                 copy_a[i] -= (copy_a[j]'*copy_a[i]) * copy_a[j]
                           if norm(copy_a[i]) < tol</pre>
```

println("Vectors are linearly dependent.")

copy_a[i] = copy_a[i]/norm(copy_a[i])

return copy_a

Out[59]: Gram_Schmidt (generic function with 1 method)

return copy_a