n [2]:	Q1 (Chelsea'srestaurant)Chelsea is a Michelin chef and he has a restaurant located at the city of Taipei. To run her Michelin restaurant, she needs 100 magic ingredients. For simplicity, her names this 100 ingredients from 1 to 100. Each ingredient corresponds to several properties, including price, quality (freshness) level, and nutrition level. And there are 100 customers in this market, with their location recorded in latitude and longitude on a 2-dimensional grid. The latitude and longitude of the customers are in the following array respectively.  Chelsea's restaurant locates at (latitude, longitude) = (0.2, 0). (red dot/trace 4)
	<pre>function kmeans(x, k; maxiters = 100, tol = 1e-5)     N = length(x) #number point of cluster     n = length(x[1]) # the dimention     distances = zeros(N) # used to store the distance of each     reps = [zeros(n) for j=1:k] # used to store representatives.     # 'assignment' is an array of N integers between 1 and k.     assignment = [ rand(1:k) for i in 1:N ] # The initial assignment is chosen randomly.     Jprevious = Inf # used in stopping condition     for iter = 1:maxiters     # Cluster j representative is average of points in cluster j.</pre>
	<pre># 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         # Show progress and terminate if J stopped decreasing.         println("Iteration ", iter, ": Jclust = ", J, ".")         if iter&gt;1&amp;&amp;abs(J-Jprevious)<tol*j assignment,="" end<="" reps="" return="" th=""></tol*j></pre>
ut[2]: n [3]:	<pre>Jprevious = J end end  kmeans (generic function with 1 method)  Random.seed!(80) X = vcat( [ [0, -0.2] + 0.1*randn(2) for i = 1:30 ],</pre>
ut[3]:	0.50
	0.00
ո [5]:	-0.500.25
	<pre>N = length(X) grps = [[X[i] for i=1:N if assignment[i] == j] for j=1:k] c1 = [[0, 0.2]] chelx = [x[1] for x in c1]; chely = [x[2] for x in c1];  Iteration 1: Jclust = 0.17042001437840845. Iteration 2: Jclust = 0.030932745932172816. Iteration 3: Jclust = 0.017346367918245634. Iteration 4: Jclust = 0.01680809465434592. Iteration 5: Jclust = 0.016747054195745868. Iteration 6: Jclust = 0.016747054195745868.</pre>
n [6]: ut[6]:	<pre>scatter([c[1] for c in grps[1]], [c[2] for c in grps[1]], markercolor = :green,label ="cust_clus_1") scatter!([c[1] for c in grps[2]], [c[2] for c in grps[2]], markercolor = :blue,label ="cust_clus_2") scatter!([c[1] for c in grps[3]], [c[2] for c in grps[3]], markercolor = :orange,label ="cust_clus_3") scatter!([c[1] for c in grps[4]], [c[2] for c in grps[4]], markercolor = :Purple,label ="cust_clus_4") scatter!(chelx ,chely, markercolor = :Red,markershape = :star4,label ="Chelsea") plot!(legend = false, grid = true, size = (500,500),</pre>
	0.25
	-0.25 -0.50 -0.75 -0.25 0.00 0.25 0.50 0.75
ո [8]։	Q2 Following question1, help Chelsea with the following problems.  since there is only one reps to match which is[0, 0.2] so the assignment should be 1 for the lenght of X
n [9]: ut[9]: [10]:	<pre>#11 Jclustl1(x,reps,assignment) =     [cityblock(x[i],reps[assignment[i]]) for i=1:length(x)] # %% codecel1 #12 Jclustl2(x,reps,assignment) =     [euclidean(x[i],reps[assignment[i]]) for i=1:length(x)]</pre> Jclustl2 (generic function with 1 method) #distance 11 12
t[10]:	dist_chel_12 = Jclustl2(x,reps,assignment)  dist_chel_11 = Jclustl1(x,reps,assignment)  100-element Vector{Float64}: 0.6037405415742244 0.2975248195514337 0.8000729267562832 0.567981247309809 0.34239099525898076 0.4998323302857391 0.35655561873714 0.6084974419928058 0.22476065902644157
	0.5012768021310703 0.4668277251532178 0.6865545740856729 0.5160176029526422 : 0.23666564134387297 0.37522160536323756 0.4095471420438517 0.2639149419491269 0.1605047590063114 0.23067100587382036 0.1766116819755589 0.3385438830965444
	0.21397081078662103 0.21171139940736902 0.2675122059191885 0.1831578966108126  twoa = mean(dist_chel_l2)  0.5182495858172949  closest distance euclidean (L2) distance  twob = smallest_distance_l2 = findmin(dist_chel_l2)
[13]: t[13]: [14]:	<pre>check_twob = euclidean([0, 0.2], X[smallest_distance_l2_argmin])</pre>
[15]: t[15]:	<pre>0.13471216806328742 what is the farthest customer to Chelsea's restaurant in terms of 1-norm?  twoc = findmax(dist_chel_l1)  (1.3849899594784718, 77) 2d Are the 5 closest customers to Chelsea's restaurant calculated using1-norm and 2-norm the same?  function small_value(x,r)</pre>
	<pre>#x = dist_chel for i=1:r</pre>
[17]: [18]:	<pre>l1_5_smallest = small_value(dist_chel_l1,5) #11 5 smallest  (0.1605047590063114, 93) (0.17103784479677606, 87) (0.1766116819755589, 93) (0.1831578966108126, 97) (0.19802513956694254, 83)  l2_5_smallest = small_value(dist_chel_l2,5) #12 5 smallest  (0.13471216806328742, 95)</pre>
	(0.1364951197145926, 99) (0.14104298008700597, 83) (0.15718976804781376, 86) (0.1602991540513241, 91)  The answer is not all is the same  Plot the predicted price on the y-axis and the true price on the x-axis. Which data point is the farthest (in terms of 2-norm) to its prediction? price! = $\beta$ quantity× qualityi + $\beta$ nutritioni + $v$ The parameters are $\beta$ quantity'(= 0.5, $\beta$ nutrition = 7.1, and $v$ = 0.1.)  Random. seed!(8) $y = \text{randn}(100)$ Random. seed!(8)
t[19]:	<pre>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</pre> 100-element Vector{Float64}: -0.000851516838157651 1.6235845696603337 2.5846052605028182 0.9414274099496751 0.8461935074630715 2.187696877681609 0.47637230397694386
	1.408735754714744 -0.21237690962130745 0.8003412498506579 0.40604368813399955 1.0406850459297647 1.7695374046991874 : 0.026392846092910094 -0.2788050730559315 1.585093458851335 2.8320382058495976 0.7505145088279714 1.3200552724372794
[20]:	<pre>0.4790338793258766 1.2879753074206888 0.9524932549190859 0.8720287887882157 1.471984940256487 0.29868172574755997  bq = 0.5 bn = 7.1 v = 0.1 pred_price = (bq * quality) + (bn * nutrition) .+ v  100-element Vector{Float64}:</pre>
	0.7173235511741521 1.0803127538307102 1.8163936703785228 0.3903592948230463 1.683788534628733 2.0733734685219045 -0.14011481117988492 1.8356227492295751 0.7359367864847383 0.8175321751660211 1.6824995751182175 -0.032165386955854136 0.8416559713167097
	0.7777161645214463 -1.9713857711282077 1.5767146273409622 2.5791730763634653 0.7259812619780989 -0.01819861318150806 0.9312579681140158 -0.045879374490436636 -0.6178279675056734 1.0770174305811908 1.1723378469696333 0.6153200637138181
[21]:	scatter(price, pred_price, label ="price")  4  3  2
[22]:	0.0 0.5 1.0 1.5 2.0 2.5  Which data point is the farthest (in terms of 2-norm) to its prediction? $ Jclustl2(x,reps) = [euclidean(x[i],reps[i]) \text{ for } i=1:length(x)] $
	<pre>x= price reps = pred_price dist_price_l2 = Jclustl2(x,reps)  100-element Vector{Float64}:</pre>
	<pre>x= price reps = pred_price dist_price_l2 = Jclustl2(x,reps)  100-element Vector{Float64}: 0.7181750680123098 0.5432718158296235 0.7682115901242954 0.5510681151266288 0.8375950271656614 0.11432340915970451 0.6164871151568287 0.42688699451483103 0.9483136961060457 0.017190925315363148 1.276455886984218 1.0728504328856188 0.9278814333824777 iii</pre>
[23]:	<pre>x= price reps = pred_price dist_price_12 = Jclust12(x,reps)  100-element Vector{Float64}: 0.7181750680123098 0.5432718158296235 0.7682115901242954 0.5510681151266288 0.8375950271656614 0.11432340915970451 0.6164871151568287 0.42688699451483103 0.9483136961060457 0.017190925315363148 1.276455886984218 1.0728504328856188</pre>
[24]: [24]: [24]: [25]:	reps = prod_price dist_price_l2 = 3clustl2(x,reps)  180-element Vector(Float64): 0.7881789680223988 0.54377181582596235 0.6164871851568287 0.6164871851568287 0.6164871851568287 0.6164871851568287 0.6164871851568287 0.62688981898787858614 0.94281898189787858614 0.942818981897878451 0.942888981897878451 0.942888981897878 0.942889781898189818 0.94281898189818 0.94281898189818 0.94281898189818 0.94281898189818 0.94281898189818 0.94281898189818 0.94281898189818 0.94281898189818 0.94281898189818 0.94281898189818 0.94288878818189898 0.94288878818189898 0.94288887881898 0.94288887881898 0.942888878818988 0.9428888888888 0.9428888888888 0.9428888888888 0.94288888888888 0.94288888888888 0.94288888888888 0.942888888888888 0.9428888888888888 0.942888888888888888888888888888888888888
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