Biostatistics HomeWork 1

${\rm KIM~SANG~HYUN}(202211545)$

2025-03-14

Contents

0.	Package and Data	2
1.	For each cancer cell line , compute average gene expression values. Identify two cell lines that have the largest and the smallest mean values. Also, include the maximum and minimum mean values	2
2.	For each gene , compute average gene expression values of 64 cancer cell lines, including "UN-KNOWN" label. Identify top 5 gene that have the largest mean expression values and top 5 genes that have smallest mean expression values . Also, include their mean values with gene ID number (1~6830)	4
3.	Suppose that group "A" contains "BREAST" and "NSCLC", group "B" has "MELANOMA", "OVARIAN" and "PROSTATE", and group "C" has "LEUKEMIA", "RENAL" and "UNKNOWN". The other 6 cell lines belong to group "D". For each cancer group, compute the mean expression values and the standard deviation of gene experssion values	5
4.	For each cancer group defined in Q3, compute the sample SD of gene expression values of individual genes. Find genes whose SD is less than 0.2 or greater than 2 for each cancer group, i.e., $SD < 0.2$ or $SD > 2$. How many genes are overlapped by 4 different cancer groups? How many genes are overlapped by exactly 3 different cancer groups? or exactly 2 different cancer groups? Also, how many genes are uniquely identified by only one cancer group? Summarize your answer, using the following table	6
5.	For each gene, compute the pairwise difference in mean expression values among 4 different cancer groups. Note that there are a total of 6 pairs among 4 cancer groups. Which gene and which cancer group pair have the largest difference in mean expression values? You should report the numerical value of the largest difference along with the gene ID number. Also, identify the corresponding cancer group pair that has the largest difference	8
6.	Only 9 different cancer cell lines have at least 2 samples. For each of these 9 cell lines, compute the pairwise distance of expression values between two samples (i, j)	s.t
	$dist(i,j) = \sqrt{\sum_{k=1}^{p} (x_{ik} - x_{jk})^2}$	
	where x_{ik} stands for the gene expression value of the <i>i</i> -th sample and the <i>k</i> -th gene, and $p=6,830$. Which cell line and two samples have the smallest pairwise distance? Include the numberical value of the smallest distance with two sample ID number $(1\sim64)$ and the name of the corresponding cancer cell line. Note that computation of distance between two samples	
	is limited the same cancer cell line	9

0. Package and Data

```
library(ISLR)
data("NCI60")
unique(NCI60$labs)
   [1] "CNS"
                      "RENAL"
                                    "BREAST"
                                                  "NSCLC"
                                                                "UNKNOWN"
## [6] "OVARIAN"
                      "MELANOMA"
                                    "PROSTATE"
                                                  "LEUKEMIA"
                                                                "K562B-repro"
## [11] "K562A-repro" "COLON"
                                    "MCF7A-repro" "MCF7D-repro"
dim(NCI60$data) # row = cancer cell lines(samples), col = gene expression
## [1]
        64 6830
str(NCI60)
## List of 2
  $ data: num [1:64, 1:6830] 0.3 0.68 0.94 0.28 0.485 ...
     ..- attr(*, "dimnames")=List of 2
    ....$ : chr [1:64] "V1" "V2" "V3" "V4" ...
    ....$ : chr [1:6830] "1" "2" "3" "4" ...
## $ labs: chr [1:64] "CNS" "CNS" "CNS" "RENAL" ...
NCI60$data[1:5, 1:10]
##
                                                    5
## V1 0.300000 1.180000 0.550000 1.140000 -0.265000 -7.000000e-02 0.350000
## V2 0.679961 1.289961 0.169961 0.379961 0.464961 5.799610e-01 0.699961
## V3 0.940000 -0.040000 -0.170000 -0.040000 -0.605000 0.000000e+00 0.090000
## V4 0.280000 -0.310000 0.680000 -0.810000 0.625000 -1.387779e-17 0.170000
## V5 0.485000 -0.465000 0.395000 0.905000 0.200000 -5.000000e-03 0.085000
##
## V1 -0.315000 -0.45000000 -0.65498050
## V2 0.724961 -0.04003899 -0.28501950
## V3 0.645000 0.43000000 0.47501950
## V4 0.245000 0.02000000 0.09501949
## V5 0.110000 0.23500000 1.49001949
sum(is.na(NCI60$data))
```

[1] 0

1. For each cancer cell line, compute average gene expression values. Identify two cell lines that have the largest and the smallest mean values. Also, include the maximum and minimum mean values.

```
sol_1 = function(x = NCI60$data, cell_name = NCI60$labs){
  avg_gene_expr_cancer_cell = apply(x, 1, mean)
  max_idx = which.max(avg_gene_expr_cancer_cell)
  min_idx = which.min(avg_gene_expr_cancer_cell)
  max cell name = cell name[max idx]
  min cell name = cell name[min idx]
  max_value = avg_gene_expr_cancer_cell[max_idx]
  min_value = avg_gene_expr_cancer_cell[min_idx]
  return(
     list(
      avg_gene_expr_cancer_cell = avg_gene_expr_cancer_cell,
      max_cell_name = max_cell_name,
      max_value = max_value,
      min_cell_name = min_cell_name,
      min_value = min_value
    )
  )
}
sol 1()
   $avg_gene_expr_cancer_cell
##
             V1
                           V2
                                         ٧3
                                                      ۷4
                                                                    V5
                                                                                  V6
##
    0.065301161
                 0.050764025
                               0.072186913
                                             0.093828598
                                                           0.148587415
                                                                        0.051984464
             ۷7
                           V8
                                         ۷9
                                                     V10
                                                                   V11
##
                 0.034898165
##
    0.045946454
                               0.030765650
                                             0.071969531
                                                           0.082987492
                                                                        0.097510887
##
            V13
                          V14
                                        V15
##
    0.043871076
                 0.080305578
                               0.086590327
                                             0.068228701
                                                           0.036918312
                                                                        0.005414269
##
                          V20
                                        V21
                                                                   V23
                               0.022486014
                                                                        0.028589256
##
    0.044108654 -0.010810123
                                             0.021204167
                                                           0.022557379
##
            V25
                          V26
                                        V27
                                                     V28
                                                                   V29
    0.055356775
                 0.076016757
                               0.037696115
                                             0.030543813
                                                           0.052745532
                                                                        0.029498474
##
##
            V31
                          V32
                                        V33
                                                     V34
                                                                   V35
                                                                                 V36
##
    0.065867052
                 0.045337045
                               0.035683095
                                             0.008683965 -0.025011443 -0.067179793
##
            V37
                          V38
                                        V39
                                                     V40
                                                                   V41
   -0.073247688 -0.059364277 -0.142086454 -0.112343414 -0.079673926
                                                                        0.016840221
##
##
            V43
                          V44
                                        V45
                                                     V46
                                                                   V47
                                                                                 V48
                               0.013481493
                                             0.008972460
                                                                        0.005350211
   -0.022744231 -0.045290706
                                                           0.009429658
            V49
                          V50
                                        V51
                                                     V52
                                                                   V53
                                                                                 V54
##
##
   -0.070572553 -0.045211715 -0.015887593
                                             0.014101316
                                                           0.050248954
                                                                       -0.033281552
##
            V55
                          V56
                                        V57
                                                     V58
                                                                   V59
   -0.043874535
                 0.016277763 -0.004496054 -0.019257425
                                                           0.040631693 0.065602038
##
            V61
                          V62
                                        V63
##
    0.040684766 0.072229801 0.021697977 0.039845104
##
## $max_cell_name
```

[1] "BREAST"

```
##
## $max_value
## V5
## 0.1485874
##
## $min_cell_name
## [1] "LEUKEMIA"
##
## $min_value
## V39
## -0.1420865
```

2. For each gene, compute average gene expression values of 64 cancer cell lines, including "UNKNOWN" label. Identify top 5 gene that have the largest mean expression values and top 5 genes that have smallest mean expression values. Also, include their mean values with gene ID number $(1\sim6830)$.

```
sol_2 = function(x = NCI60$data){
   avg_gene_expr_gene = apply(x ,2, mean)

   top_5_max_idx = order(avg_gene_expr_gene, decreasing = TRUE)[1 : 5]
   top_5_min_idx = order(avg_gene_expr_gene)[1 : 5]

   top_5_max_gene = avg_gene_expr_gene[top_5_max_idx]
   top_5_min_gene = avg_gene_expr_gene[top_5_min_idx]

return(
   list(
        top_5_max_gene = top_5_max_gene,
        top_5_min_gene = top_5_min_gene
   )
)
}
sol_2()
```

3. Suppose that group "A" contains "BREAST" and "NSCLC", group "B" has "MELANOMA", "OVARIAN" and "PROSTATE", and group "C" has "LEUKEMIA", "RENAL" and "UNKNOWN". The other 6 cell lines belong to group "D". For each cancer group, compute the mean expression values and the standard deviation of gene experssion values.

```
## $group_mean
## A 0.025053741
## B 0.040698588
## C 0.008992817
## D 0.005848171
##
## $group_sd
## A 0.8006589
## B 0.7368874
## C 0.8443378
## D 0.7924379
```

4. For each cancer group defined in Q3, compute the sample SD of gene expression values of individual genes. Find genes whose SD is less than 0.2 or greater than 2 for each cancer group, i.e., SD < 0.2 or SD > 2. How many genes are overlapped by 4 different cancer groups? How many genes are overlapped by exactly 3 different cancer groups? or exactly 2 different cancer groups? Also, how many genes are uniquely identified by only one cancer group? Summarize your answer, using the following table.

```
sol_4 = function(){
 group_sd = lapply(groups,
                  function(grp) {
                    apply(NCI60$data[cell_group == grp, ], 2, sd)
  names(group_sd) = c("A", "B", "C", "D")
  filtered_genes = lapply(group_sd,
                        function(sd_value){
                          names(sd value[sd value < 0.2 | sd value > 2])
  unique_genes = unique(c(filtered_genes[["A"]],
                            filtered genes[["B"]],
                            filtered_genes[["C"]],
                            filtered_genes[["D"]]))
  num_overlapped = factor(apply(sapply(filtered_genes,
                                       function(g_list) {
                                         unique_genes %in% g_list
                                         \}), 1, sum), levels = 1:4)
  sol_df = data.frame(
   x4 = length(num_overlapped[num_overlapped == "4"]),
   x3 = length(num_overlapped[num_overlapped == "3"]),
   x2 = length(num_overlapped[num_overlapped == "2"]),
   x1 = length(num_overlapped[num_overlapped == "1"])
  )
  rownames(sol_df) = "The number of genes"
  colnames(sol df) = c("4 groups", "3 groups", "2 groups", "1 groups")
 return(
   list(
      filtered_genes = filtered_genes,
      sol_df = sol_df
      ))
}
sol_4()
```

^{## \$}filtered_genes
\$filtered_genes\$A

```
[1] "16"
               "111" "112" "113" "134" "196" "243" "245" "248" "251"
##
    [11] "252" "256" "257" "266" "267" "273" "281" "286" "472" "975"
##
    [21] "1106" "1215" "1258" "1865" "2068" "2504" "2838" "2875" "2914" "2927"
##
    [31] "3320" "3383" "3438" "3518" "3525" "3543" "3936" "3956" "3957" "4050"
##
    [41] "4154" "4280" "4288" "4344" "4353" "4354" "4699" "4700" "4701" "5036"
##
    [51] "5142" "5221" "5275" "5276" "5353" "5476" "5477" "5555" "5556" "5557"
##
    [61] "5586" "5587" "5661" "5692" "5705" "5706" "5707" "5723" "5732" "5760"
    [71] "5803" "5804" "5805" "5828" "5829" "5838" "5843" "5845" "5913" "5940"
##
    [81] "5942" "5943" "5948" "5980" "6128" "6148" "6149" "6150" "6151" "6152"
##
    [91] "6153" "6156" "6157" "6263" "6264" "6268" "6277" "6278" "6279" "6321"
##
   [101] "6328" "6356" "6391" "6392" "6393" "6415" "6416" "6419" "6429" "6430"
   [111] "6453" "6564" "6612" "6614" "6615" "6616" "6622" "6718"
##
##
  $filtered_genes$B
   [1] "124" "125"
                     "128" "130"
                                   "133" "134" "196" "241"
                                                               "242"
   [11] "252" "256"
                     "257" "286"
                                   "287" "408" "416" "561" "580" "581"
##
   [21] "592" "754" "755" "770" "1067" "1110" "1508" "1664" "1888" "1896"
   [31] "1897" "2100" "2216" "2239" "2551" "2678" "2680" "2891" "3234" "3706"
   [41] "3713" "3957" "4093" "4094" "4280" "4288" "4289" "4304" "4306" "4308"
   [51] "4320" "4327" "4344" "4353" "4354" "4375" "4383" "4387" "4388" "4425"
##
   [61] "4426" "4699" "4700" "4701" "4716" "4971" "5094" "5275" "5276" "5353"
## [71] "5555" "5556" "5557" "5586" "5804" "5948" "6149" "6150" "6322" "6356"
## [81] "6391" "6392" "6393" "6434" "6554" "6635" "6710"
##
##
  $filtered_genes$C
                             "187" "196" "243"
                                                  "252"
                                                          "256" "281" "286"
     [1] "16"
               "78"
                       "133"
##
    [11] "301" "415" "515" "707" "754" "755" "756"
                                                         "806" "1199" "1229"
    [21] "1387" "1388" "1389" "1390" "1391" "2068" "2070" "2074" "2080" "2081"
    [31] "2082" "2083" "2102" "3234" "3248" "3282" "3372" "3373" "3490" "3491"
##
    [41] "3518" "3525" "3894" "4085" "4131" "4154" "4245" "4344" "4354" "4699"
    [51] "4700" "4701" "4716" "5127" "5221" "5270" "5301" "5336" "5392" "5481"
##
    [61] "5489" "5496" "5506" "5510" "5586" "5587" "5588" "5705" "5712" "5721"
##
    [71] "5729" "5732" "5758" "5760" "5774" "5796" "5803" "5804" "5805" "5867"
##
    [81] "5868" "5869" "5870" "5872" "5878" "5884" "5899" "5902" "5910" "5917"
##
    [91] "5921" "5927" "5928" "5937" "5940" "5941" "5942" "5943" "5946" "5948"
##
   [101] "5950" "5962" "5972" "5973" "5976" "5979" "5980" "5981" "5993" "6009"
   [111] "6010" "6017" "6018" "6035" "6039" "6046" "6084" "6085" "6086" "6087"
   [121] "6124" "6148" "6149" "6150" "6151" "6152" "6153" "6154" "6156" "6157"
   [131] "6169" "6243" "6272" "6274" "6277" "6278" "6279" "6288" "6289" "6382"
   [141] "6391" "6392" "6393" "6412" "6413" "6414" "6415" "6416" "6429" "6430"
   [151] "6592" "6596" "6635" "6644" "6646" "6688" "6689" "6710" "6717" "6817"
##
   $filtered_genes$D
##
##
     [1] "16"
              "112" "113" "161" "188" "224" "227"
                                                         "228" "229" "243"
    [11] "248" "252" "256" "257" "267" "286" "301" "412" "582" "707"
    [21] "716" "754" "755" "770" "975" "1187" "1380" "1382" "1388" "1389"
##
    [31] "1390" "1391" "1393" "1396" "1613" "1716" "2080" "2081" "2096" "2104"
##
    [41] "2302" "3212" "3424" "3936" "3957" "4010" "4057" "4060" "4093" "4094"
##
    [51] "4119" "4154" "4231" "4344" "4472" "4612" "4644" "4699" "4700" "4701"
    [61] "4703" "4704" "4706" "4845" "4994" "5031" "5142" "5472" "5481" "5646"
##
    [71] "5680" "5691" "5692" "5696" "5705" "5706" "5707" "5732" "5804" "5805"
##
    [81] "5838" "5867" "5868" "5869" "5870" "5916" "5917" "5937" "5948" "5980"
##
    [91] "6068" "6149" "6274" "6391" "6392" "6393" "6415" "6416" "6612" "6635"
##
## [101] "6646" "6687" "6688" "6689"
```

5. For each gene, compute the pairwise difference in mean expression values among 4 different cancer groups. Note that there are a total of 6 pairs among 4 cancer groups. Which gene and which cancer group pair have the largest difference in mean expression values? You should report the numerical value of the largest difference along with the gene ID number. Also, identify the corresponding cancer group pair that has the largest difference.

6. Only 9 different cancer cell lines have at least 2 samples. For each of these 9 cell lines, compute the pairwise distance of expression values between two samples (i, j) s.t

$$dist(i,j) = \sqrt{\sum_{k=1}^{p} \left(x_{ik} - x_{jk}\right)^2}$$

where x_{ik} stands for the gene expression value of the *i*-th sample and the *k*-th gene, and p=6,830. Which cell line and two samples have the smallest pairwise distance? Include the nummerical value of the smallest distance with two sample ID number (1 ~ 64) and the name of the corresponding cancer cell line. Note that computation of distance between two samples is limited the same cancer cell line.

```
filter_cell = names(table(NCI60$labs)[table(NCI60$labs) >= 2])
compr = Inf
ans_samples = NULL
ans_cell = NULL
for(cell in filter_cell){
  cell_id = which(NCI60$labs == cell)
  if(length(cell_id) >= 2){
    for(i in 1:(length(cell_id) - 1)){
      for(j in (i + 1):length(cell_id)){
        dist_cell = sqrt(sum((NCI60$data[cell_id[i], ] - NCI60$data[cell_id[j], ])^2))
        if(dist_cell < compr){</pre>
          compr = dist_cell
          ans_samples = c(cell_id[i], cell_id[j])
          ans_cell = cell
        }
      }
    }
 }
cat("Smallest dist:", compr, "\n",
    "Sample ID:", ans_samples, "\n",
    "Name of cancer cell line:", ans_cell)
```

```
## Smallest dist: 39.10562
## Sample ID: 57 58
## Name of cancer cell line: BREAST
```