

# Lab6\_1

2024-05-07

## Section 1-3:

```
library(ggplot2)
```

```
## Warning: package 'ggplot2' was built under R version 4.3.2
```

```
data.df <- read.csv("phagocyte_PCA.csv", row.names=1)
# 5
gene.PCA <- prcomp( t(data.df) )
# 6
PCA.summary <- summary(gene.PCA)
PCA.summary
```

```
## Importance of components:
##
```

	PC1	PC2	PC3	PC4	PC5
## Standard deviation	880.6774	520.3121	227.79013	201.67594	193.66738
## Proportion of Variance	0.6303	0.2200	0.04217	0.03305	0.03048
## Cumulative Proportion	0.6303	0.8503	0.89243	0.92548	0.95596

```
##
```

	PC6	PC7	PC8	PC9	PC10	PC11
## Standard deviation	150.26078	102.12845	84.35153	70.53612	50.99879	34.89438
## Proportion of Variance	0.01835	0.00848	0.00578	0.00404	0.00211	0.00099
## Cumulative Proportion	0.97431	0.98279	0.98857	0.99261	0.99472	0.99571

```
##
```

	PC12	PC13	PC14	PC15	PC16	PC17
## Standard deviation	32.69040	29.91605	27.2092	23.79487	21.48151	18.74291
## Proportion of Variance	0.00087	0.00073	0.0006	0.00046	0.00037	0.00029
## Cumulative Proportion	0.99658	0.99731	0.9979	0.99837	0.99875	0.99903

```
##
```

	PC18	PC19	PC20	PC21	PC22	PC23
## Standard deviation	17.61497	16.26246	14.86637	13.85835	11.2509	8.76728
## Proportion of Variance	0.00025	0.00021	0.00018	0.00016	0.0001	0.00006
## Cumulative Proportion	0.99928	0.99950	0.99968	0.99983	0.9999	1.00000

```
##
## Standard deviation    1.146e-13
## Proportion of Variance 0.000e+00
## Cumulative Proportion 1.000e+00
```

```
# 7
PCA.importance <- PCA.summary$importance
PCA.importance
```

```
##          PC1      PC2      PC3      PC4      PC5
## Standard deviation  880.67743 520.31210 227.79013 201.67594 193.66738
## Proportion of Variance  0.63027  0.22000  0.04217  0.03305  0.03048
## Cumulative Proportion  0.63027  0.85027  0.89243  0.92548  0.95596
##          PC6      PC7      PC8      PC9      PC10      PC11
## Standard deviation  150.26078 102.12845 84.35153 70.53612 50.99879 34.89438
## Proportion of Variance  0.01835  0.00848  0.00578  0.00404  0.00211  0.00099
## Cumulative Proportion  0.97431  0.98279  0.98857  0.99261  0.99472  0.99571
##          PC12      PC13      PC14      PC15      PC16      PC17
## Standard deviation  32.69040 29.91605 27.20925 23.79487 21.48151 18.74291
## Proportion of Variance  0.00087  0.00073  0.00060  0.00046  0.00037  0.00029
## Cumulative Proportion  0.99658  0.99731  0.99791  0.99837  0.99875  0.99903
##          PC18      PC19      PC20      PC21      PC22      PC23
## Standard deviation  17.61497 16.26246 14.86637 13.85835 11.25089 8.767283
## Proportion of Variance  0.00025  0.00021  0.00018  0.00016  0.00010  0.000060
## Cumulative Proportion  0.99928  0.99950  0.99968  0.99983  0.99994  1.000000
##          PC24
## Standard deviation  1.146105e-13
## Proportion of Variance  0.000000e+00
## Cumulative Proportion  1.000000e+00
```

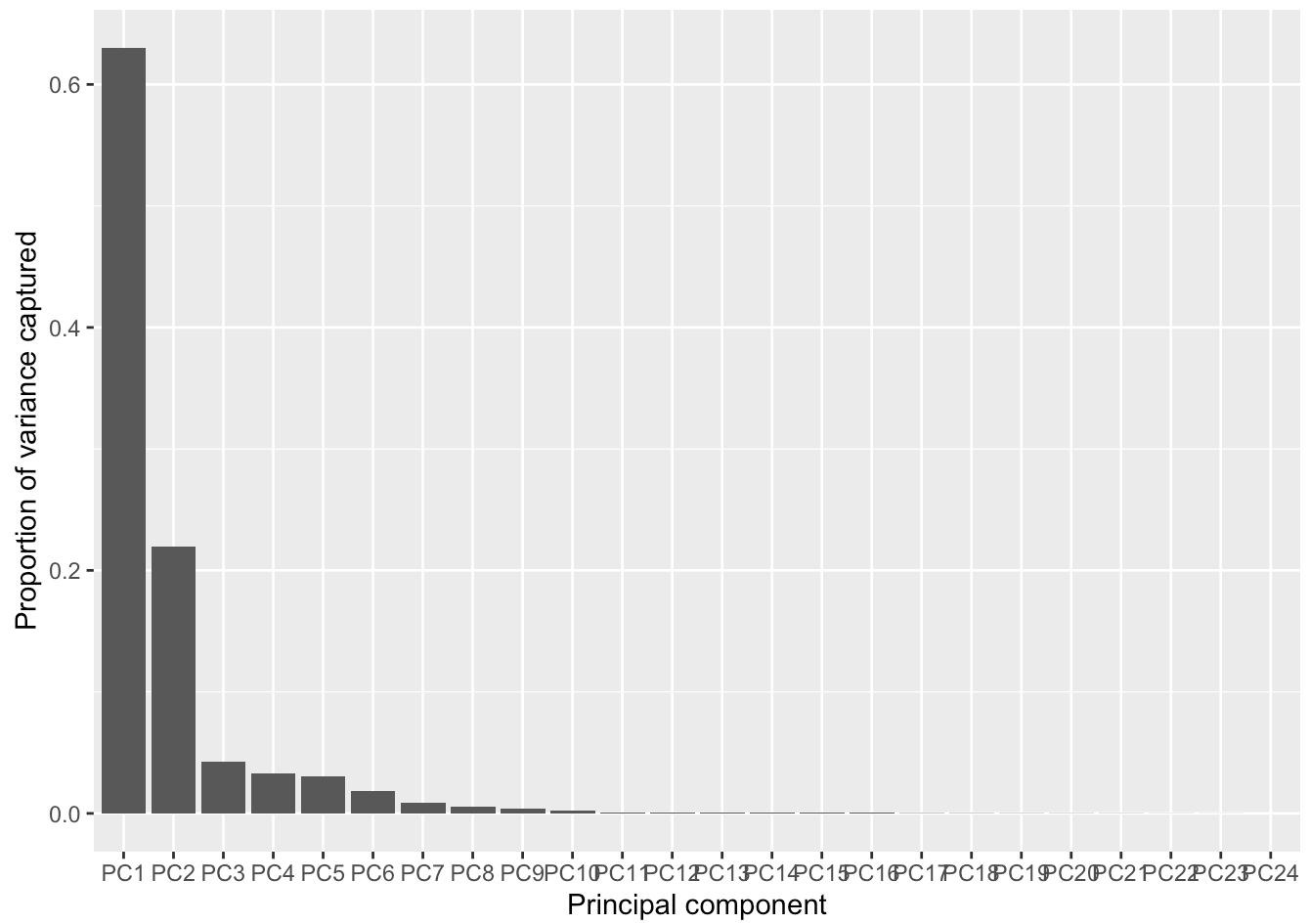
```
# 8
prop.var <- PCA.importance[2,]
prop.var
```

```
##      PC1      PC2      PC3      PC4      PC5      PC6      PC7      PC8      PC9      PC10
## 0.63027 0.22000 0.04217 0.03305 0.03048 0.01835 0.00848 0.00578 0.00404 0.00211
##      PC11      PC12      PC13      PC14      PC15      PC16      PC17      PC18      PC19      PC20
## 0.00099 0.00087 0.00073 0.00060 0.00046 0.00037 0.00029 0.00025 0.00021 0.00018
##      PC21      PC22      PC23      PC24
## 0.00016 0.00010 0.00006 0.00000
```

```
# 9
scree.df <- data.frame( prop.var )
scree.df
```

```
##      prop.var
## PC1    0.63027
## PC2    0.22000
## PC3    0.04217
## PC4    0.03305
## PC5    0.03048
## PC6    0.01835
## PC7    0.00848
## PC8    0.00578
## PC9    0.00404
## PC10   0.00211
## PC11   0.00099
## PC12   0.00087
## PC13   0.00073
## PC14   0.00060
## PC15   0.00046
## PC16   0.00037
## PC17   0.00029
## PC18   0.00025
## PC19   0.00021
## PC20   0.00018
## PC21   0.00016
## PC22   0.00010
## PC23   0.00006
## PC24   0.00000
```

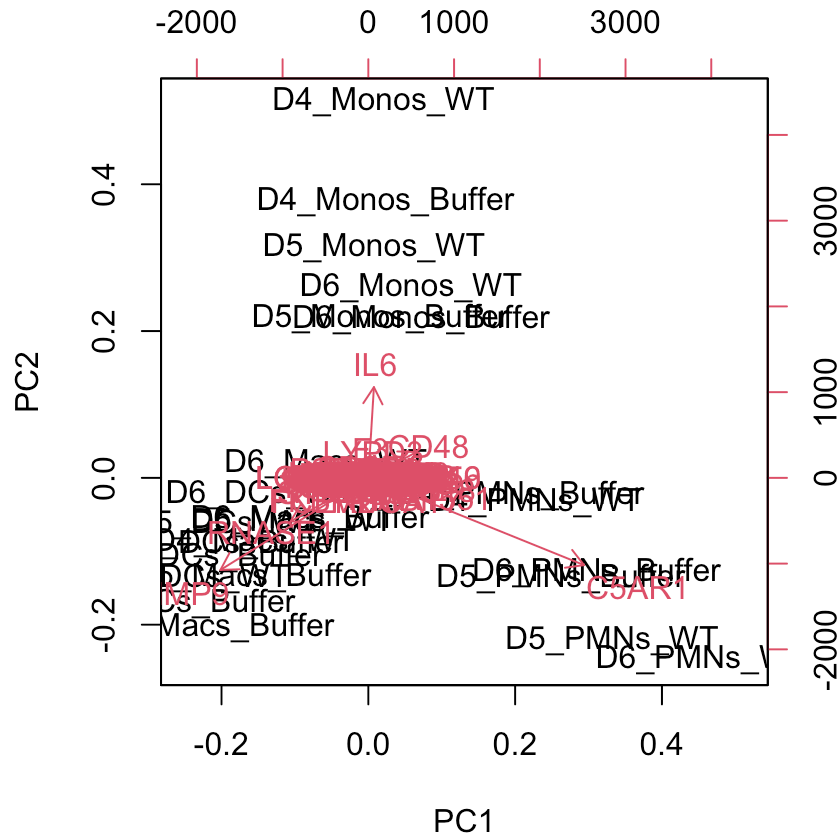
```
# 10
PCs <- factor(rownames(scree.df), levels=unique(rownames(scree.df)))
ggplot(data=scree.df, aes(x = PCs, y = prop.var)) +
  geom_col()+
  ylab("Proportion of variance captured") +
  xlab("Principal component")
```



```
# 11  
biplot(gene.PCA)
```

[illegible]

```
#12
PCA.df <- data.frame( gene.PCA$x )
PCA.df
```



##		PC1	PC2	PC3	PC4	PC5
##	D4_DCs_Buffer	-1086.26059	-434.34372	-168.0918007	-23.050104	76.363293
##	D4_Macs_Buffer	-898.77388	-516.92406	-380.8469217	238.732577	39.881883
##	D4_Monos_Buffer	101.95483	962.96582	-170.4603292	-75.480326	-4.691244
##	D4_PMNs_Buffer	891.46565	-59.22977	45.4747264	-89.103375	-432.498565
##	D5_DCs_Buffer	-926.00540	-274.03063	217.3227077	-371.650618	73.852051
##	D5_Macs_Buffer	-682.92173	-344.11325	-167.0428060	231.559060	-114.746006
##	D5_Monos_Buffer	71.74558	553.58670	290.9172466	123.187035	-143.008802
##	D5_PMNs_Buffer	1131.20191	-346.26804	10.9046835	-130.003762	-225.621096
##	D6_DCs_Buffer	-804.41153	-229.47746	181.0590297	-277.395019	165.734514
##	D6_Macs_Buffer	-342.63183	-142.57482	-49.1611805	286.553117	-13.503552
##	D6_Monos_Buffer	310.70559	552.82570	315.7150464	265.978773	182.412389
##	D6_PMNs_Buffer	1341.43290	-325.74027	42.0285344	11.564080	164.992866
##	D4_DCs_WT	-1018.14230	-348.94226	-132.1399337	-8.795725	22.982257
##	D4_Macs_WT	-667.19622	-222.64829	-175.5614787	127.337528	-14.563627
##	D4_Monos_WT	88.62317	1309.02100	-621.3922255	-306.025740	114.694912
##	D4_PMNs_WT	978.01649	-85.12709	-9.7001432	-57.798207	-321.233817
##	D5_DCs_WT	-833.13604	-161.04936	251.4749683	-292.860025	20.461886
##	D5_Macs_WT	-446.25367	-154.14439	0.8810619	170.485080	-137.723448
##	D5_Monos_WT	32.97787	799.26038	93.1387746	50.771375	-60.643269
##	D5_PMNs_WT	1432.08717	-563.29412	-117.5883976	-153.150056	-119.041228
##	D6_DCs_WT	-637.31264	-55.05409	268.5646535	-220.338769	83.726360
##	D6_Macs_WT	-250.50647	52.53769	97.7952249	230.934077	-90.356849
##	D6_Monos_WT	250.23376	661.03563	281.9142365	230.385482	204.641782
##	D6_PMNs_WT	1963.10737	-628.27129	-105.2056778	38.163542	527.887309
##		PC6	PC7	PC8	PC9	PC10
##	D4_DCs_Buffer	-162.068628	10.99440	-30.0232081	110.703008	62.556833
##	D4_Macs_Buffer	-149.707372	14.87713	-118.4201739	-108.900655	-13.475798
##	D4_Monos_Buffer	-26.054489	-23.47373	48.6883665	-16.918790	35.859717
##	D4_PMNs_Buffer	-203.427811	-64.77095	61.5563731	-60.462008	54.696911
##	D5_DCs_Buffer	145.294802	-140.59217	-68.3337411	-108.295900	5.041853
##	D5_Macs_Buffer	298.317609	-135.78896	57.8686956	115.685844	19.259934
##	D5_Monos_Buffer	39.882900	97.61020	-144.8377048	35.256999	-34.191283
##	D5_PMNs_Buffer	153.111995	170.32012	-71.6229702	10.548275	86.553026
##	D6_DCs_Buffer	-151.231374	153.97080	92.4891284	57.905864	-30.444593
##	D6_Macs_Buffer	72.075621	59.18230	121.2558591	-21.234314	-45.100518
##	D6_Monos_Buffer	-79.288078	-45.72177	7.2629360	-8.800839	43.432307
##	D6_PMNs_Buffer	65.900801	61.59926	36.4950582	-52.815738	113.619415
##	D4_DCs_WT	-150.418036	-45.92233	-37.5112271	119.792231	52.039339
##	D4_Macs_WT	-121.706001	67.78325	-85.9992905	-133.025636	-17.929902
##	D4_Monos_WT	93.988001	14.31380	51.7995660	-11.718114	-5.494418
##	D4_PMNs_WT	-272.772931	-202.57806	58.5634169	22.357992	-73.363070
##	D5_DCs_WT	184.304429	-140.34040	-32.4586405	-60.099600	-9.678785
##	D5_Macs_WT	265.261744	-26.11149	65.0408203	7.616995	-23.937829
##	D5_Monos_WT	29.106734	52.86305	-160.3610593	58.010071	-58.721577
##	D5_PMNs_WT	90.069960	64.50520	-73.6328523	71.240667	-70.701234
##	D6_DCs_WT	-103.275907	128.51150	110.8579666	25.563024	-35.230309
##	D6_Macs_WT	58.756920	119.32535	139.1246635	-83.517573	-28.496953
##	D6_Monos_WT	-81.324181	-115.05314	-28.5281880	28.475830	35.225117
##	D6_PMNs_WT	5.203293	-75.50338	0.7262056	2.632368	-61.518183
##		PC11	PC12	PC13	PC14	PC15
##	D4_DCs_Buffer	-12.349863	-14.9872267	0.6479082	-6.097794	29.1080196

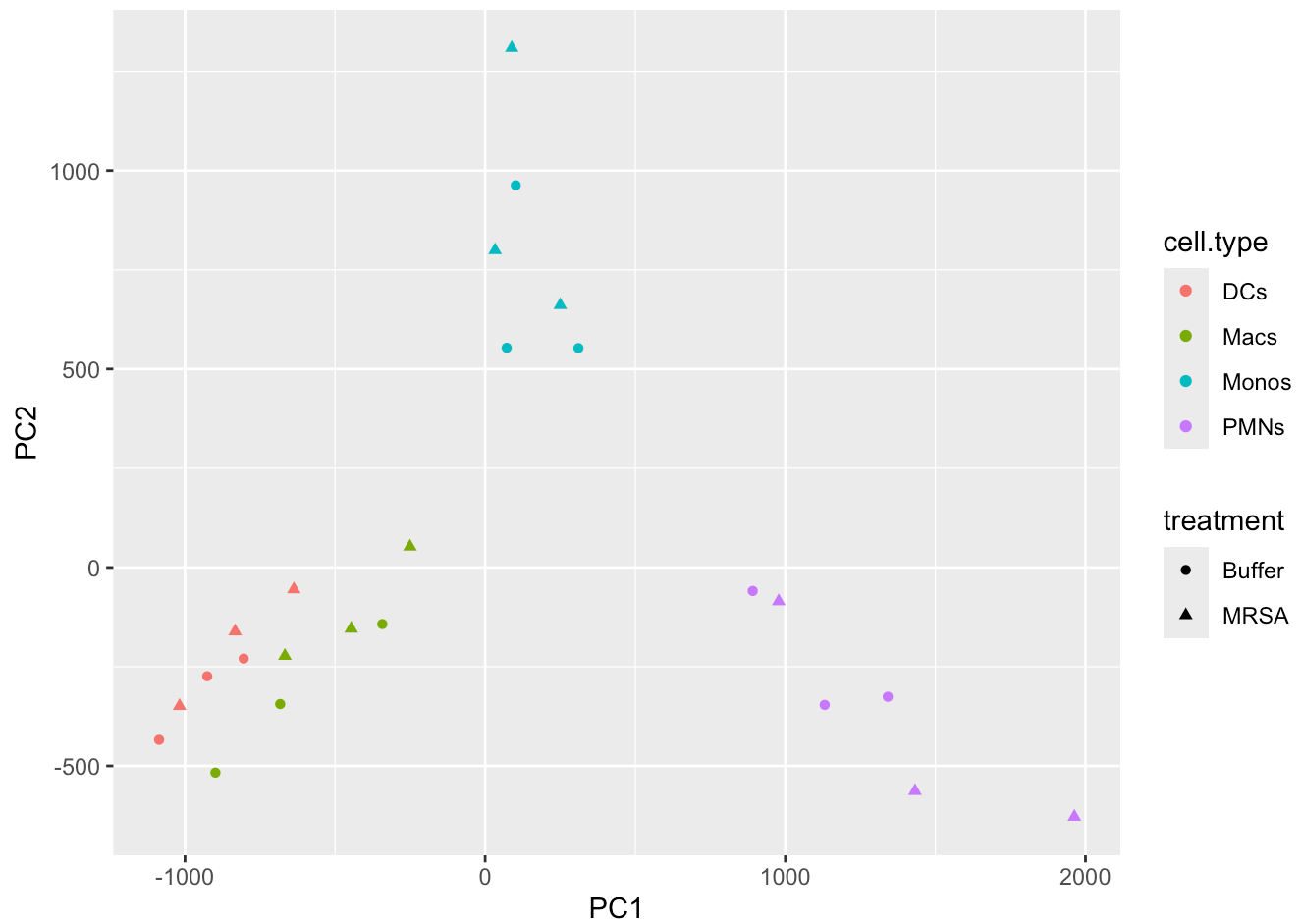
##	D4_Macs_Buffer	-17.371754	34.2653688	25.9063928	30.018218	-2.4230377
##	D4_Monos_Buffer	27.544216	-40.1875166	4.4841942	14.103019	48.2136322
##	D4_PMNs_Buffer	-78.143257	0.8307577	-34.7386751	35.712532	-27.0689306
##	D5_DCs_Buffer	2.965369	18.7389204	-0.8826358	-1.818678	42.5296043
##	D5_Macs_Buffer	-27.019604	10.8345129	44.3206713	18.132246	0.6131858
##	D5_Monos_Buffer	-19.902901	-77.2955939	20.9673936	24.808682	13.3426718
##	D5_PMNs_Buffer	2.310111	28.3802560	-39.1301624	-16.510451	32.2427573
##	D6_DCs_Buffer	-30.186519	40.6981438	26.0205384	22.564863	9.2853386
##	D6_Macs_Buffer	-2.888303	15.3496646	-34.7065990	5.465610	27.8298005
##	D6_Monos_Buffer	44.103309	2.3565281	1.9095178	62.848631	-9.1770641
##	D6_PMNs_Buffer	4.821081	-11.4770156	78.8544430	-32.126230	-23.7106498
##	D4_DCs_WT	37.639727	-42.4759787	-47.7602161	-25.379978	-29.8674295
##	D4_Macs_WT	29.080421	-15.3509399	-1.2286008	-20.031317	-13.8651145
##	D4_Monos_WT	-3.068728	10.3900550	-4.7999037	3.710327	-21.7936431
##	D4_PMNs_WT	29.946657	0.1403025	41.5247012	-36.775344	21.1932618
##	D5_DCs_WT	8.668070	-24.7316388	-11.7544655	1.651255	-34.5232886
##	D5_Macs_WT	-8.837224	-3.6276731	-13.0823739	-6.991436	-16.6809013
##	D5_Monos_WT	-55.538175	28.4505718	12.4637680	-43.743638	-13.4869175
##	D5_PMNs_WT	81.536737	29.6390978	-2.2937708	38.913492	-16.7490867
##	D6_DCs_WT	1.392085	-8.1569410	8.5710851	-3.536407	-22.7960030
##	D6_Macs_WT	23.786949	-17.2101709	-14.3064662	-35.983619	1.9099007
##	D6_Monos_WT	14.627637	76.1776218	-28.3671412	-25.807565	-4.0151368
##	D6_PMNs_WT	-53.116042	-40.7511060	-32.6196031	-3.126418	9.8890309
##		PC16	PC17	PC18	PC19	PC20
##	D4_DCs_Buffer	14.8881315	-17.136942	8.1040069	8.95491331	-36.994590
##	D4_Macs_Buffer	-18.3672541	-4.408282	-14.0848086	21.62946350	9.234952
##	D4_Monos_Buffer	25.9863659	-31.564351	-35.5819664	0.06706518	16.144401
##	D4_PMNs_Buffer	24.9785086	6.619780	-4.0977343	-5.74585468	-4.044646
##	D5_DCs_Buffer	24.6584239	25.683368	19.8777541	-9.33877030	-7.808993
##	D5_Macs_Buffer	-8.0749589	11.543494	-12.9069604	-35.55308501	9.580811
##	D5_Monos_Buffer	-1.1441016	30.965649	11.6477572	14.74475976	13.885008
##	D5_PMNs_Buffer	-47.5300705	-11.508049	4.4389717	-6.53343478	9.592618
##	D6_DCs_Buffer	1.3869117	-6.555884	9.1164739	1.24413017	25.107403
##	D6_Macs_Buffer	-10.4178292	31.720944	-15.4966896	22.60106528	-18.446626
##	D6_Monos_Buffer	-28.1307256	-14.794853	22.5308176	-14.68742868	-20.269236
##	D6_PMNs_Buffer	13.7848260	12.935422	-2.6090602	17.04812177	-5.043624
##	D4_DCs_WT	-0.9978543	21.509119	0.6884525	1.90799869	14.847016
##	D4_Macs_WT	10.2376619	-8.323051	12.5345736	-35.74050154	9.076861
##	D4_Monos_WT	-16.5376095	19.489329	21.2503898	2.78200908	-5.106050
##	D4_PMNs_WT	-30.4088483	-1.900497	13.6957238	6.24807921	1.098943
##	D5_DCs_WT	-32.4656220	-17.866655	-31.6308891	16.77904315	-4.689603
##	D5_Macs_WT	23.2850045	-41.361594	41.3084926	26.55977772	11.437460
##	D5_Monos_WT	6.7108809	-16.138329	-13.4250782	-9.90866432	-21.566768
##	D5_PMNs_WT	35.8828309	3.352030	-11.5652113	1.31787008	-8.291938
##	D6_DCs_WT	-10.3996000	-1.029277	-11.0810509	-10.89463843	-4.369804
##	D6_Macs_WT	8.4330072	4.541663	-6.2048537	-13.08790599	-7.772732
##	D6_Monos_WT	17.5196920	11.038688	-7.4245751	8.93873266	19.743973
##	D6_PMNs_WT	-3.2777712	-6.811718	0.9154643	-9.33274586	4.655166
##		PC21	PC22	PC23	PC24	
##	D4_DCs_Buffer	-24.2930543	-10.08123641	2.12864100	4.955246e-13	
##	D4_Macs_Buffer	-6.4211308	19.55460403	-3.88069651	-1.214935e-14	
##	D4_Monos_Buffer	11.0132977	2.60867696	-3.49389213	1.076509e-13	



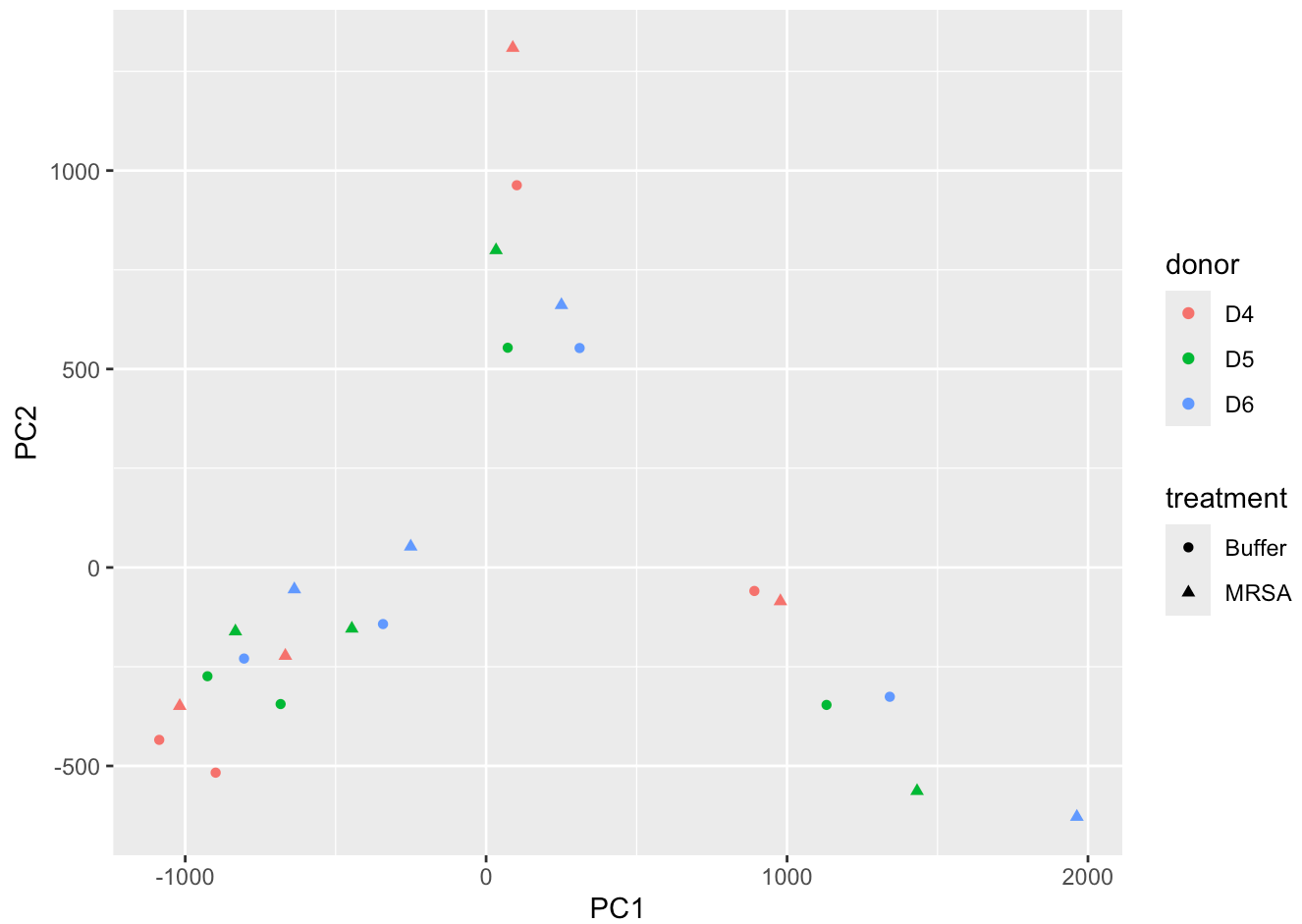
```
## D4_PMNBuffer -0.3885125 -0.82273607 0.97366790 -3.323626e-13
## D5_DCBuffer 8.1038533 12.57879223 -3.41394617 3.201490e-13
## D5_MacBuffer -5.0570293 -2.11651602 -1.01477836 1.404163e-13
## D5_MonoBuffer -14.7241470 -4.70727320 -0.28844870 -2.378440e-13
## D5_PMNBuffer -2.3675790 1.12393636 -2.83253396 -1.212745e-13
## D6_DCBuffer 10.1050797 -6.48169437 18.02629781 2.260518e-14
## D6_MacBuffer 24.9335751 -17.70935886 -3.91355816 8.540477e-15
## D6_MonoBuffer 12.5793884 6.29854458 3.24511874 -1.307461e-13
## D6_PMNBuffer 11.0570920 -2.77627285 0.16969309 -1.113510e-13
## D4_DCWT 22.6642543 14.12935878 3.34576167 5.205975e-13
## D4_MacWT 4.2207021 -28.39446669 -2.41958642 9.687997e-14
## D4_MonoWT -9.4637524 -1.65849919 1.34595114 -1.101588e-12
## D4_PMNWT 0.5059784 -0.98760457 0.21897722 1.987633e-13
## D5_DCWT -7.8882684 -13.12819625 9.78422490 9.867454e-14
## D5_MacWT 5.0329652 2.08791332 -7.19506344 -1.742638e-13
## D5_MonoWT 22.2808317 7.65963778 2.98713156 8.809040e-14
## D5_PMNWT -4.4408967 0.09292492 1.74616172 -1.805075e-13
## D6_DCWT -9.2075109 7.51273232 -29.35189184 2.055847e-13
## D6_MacWT -22.9352320 21.16168262 17.58760173 -1.623109e-13
## D6_MonoWT -22.4107094 -8.62190790 -3.78181595 -5.466270e-13
## D6_PMNWT -2.8991950 2.67695850 0.02698314 7.456405e-13
```

```
# 13
donor <- rep(c(rep("D4", 4), rep("D5", 4), rep("D6",4)), 2)
cell.type <- rep(c("DCs", "Mac", "Mono", "PMNs"), 6)
treatment <- c(rep("Buffer", 12), rep("MRSA", 12))

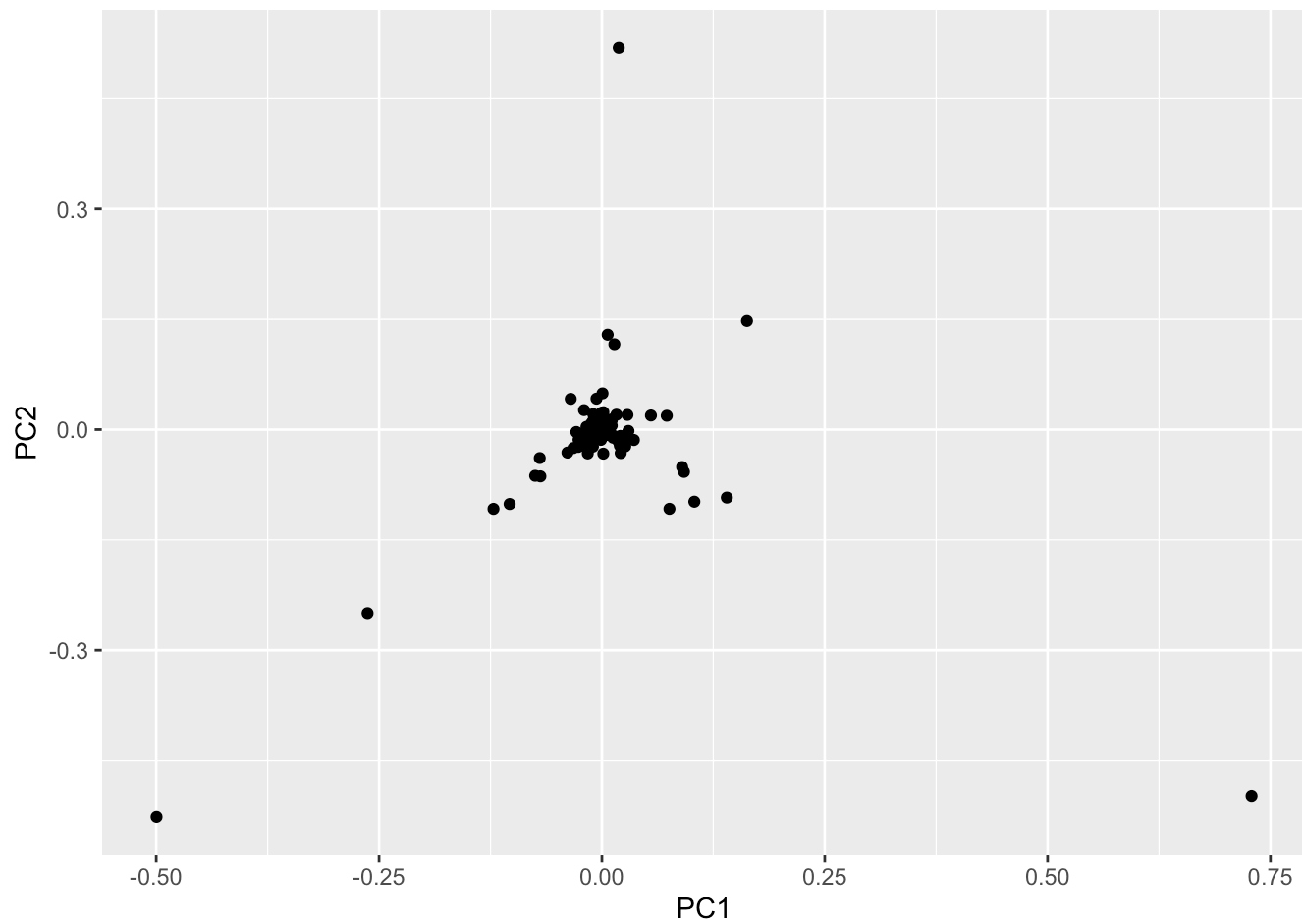
# 14
PC1 <- PCA.df$PC1
PC2 <- PCA.df$PC2
ggplot(data=PCA.df, aes(x = PC1, y = PC2, colour=cell.type, shape=treatment)) +
  geom_point()+
  ylab("PC2") +
  xlab("PC1")
```



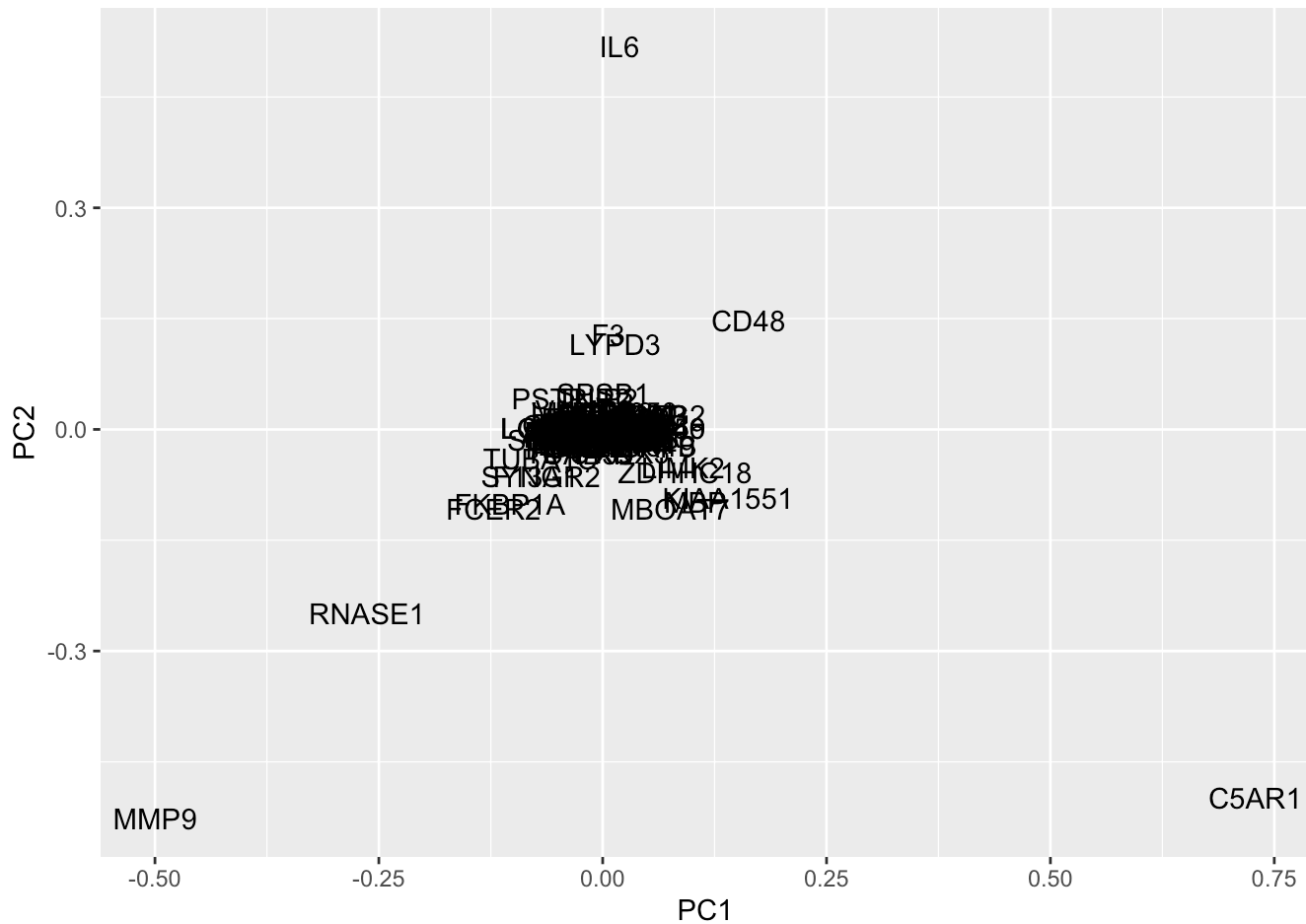
```
# 15
ggplot(data=PCA.df, aes(x = PC1, y = PC2, colour=donor, shape=treatment)) +
  geom_point()+
  ylab("PC2") +
  xlab("PC1")
```



```
# 16
rotation <- data.frame(gene.PCA$rotation)
ggplot(data=rotation, aes(x = PC1, y = PC2)) +
  geom_point()+
  ylab("PC2") +
  xlab("PC1")
```



```
ggplot(rotation, aes(x = PC1, y = PC2)) +  
  geom_text( label=row.names(rotation) )+  
  ylab("PC2") +  
  xlab("PC1")
```



## Section 4: Assessment

### Q1:

PC1 capture 0.63027, PC2 capture 0.22, so first two components capture  $0.63027 + 0.22 = 0.85027$ .

### Q2:

The minimum number of components is 4 as  $0.63027 + 0.22 + 0.04217 + 0.03305 > 0.9$

### Q3:

Too much overlapping, can't see the label very clear.

### Q4:

Cell type is most distinctly clustered by PCA as we can see from the plot that each different cell type are most spatially grouped by their category. This indicate the there is a strong correlation between gene expression patterns and the categories represented by the cell types. It means that gene expression of the samples within each category in cell type are very similar to each other and they are distinct from gene expression in the other categories.

**Q5:**

C5AR1 has the most weight in PC1 and IL6 and MMP9 have the most weight in PC2 as they are very far very the other data point.

**Q6:**

We can see the mystery data point is cluster into the group of Monos, so it is likely to be Monos.

**Q7:**

In our PCA, we only use about 250 randomly selected genes, but in the figure 1B, they used 1000 genes. So their genes selected for PCA are likely to capture a broader range of variation across samples, leading to more distinct clusters. So they have more support evident to ensure their cluster and make a clear cluster and the group are more distinct.

**Q8:**

This paper seek answer for the question that how does *Staphylococcus aureus* suppress the human immune response during infection.

**Q9:**

These 2 additional treatment group work as control group, it can be used to compare the result with the cell infected with the MRSA to see if those cells can get a similar response. These groups allows the author to determine if the gene expression changes were specific to MRSA infection or from a general inflammatory response.

**Q10:**

Figure 3E provided the evident in patients.

**Q11:**

It is in the GEO database: gene expression data from myeloid cells and whole blood exposed to *S. aureus* have been deposited in GEO under accession number GSE193219. The cytokine profile studies are publicly available (accession no. GSE131990). Clinical cohort data were obtained from publicly available studies (accession nos. GSE40396 and GSE30119).