lab6

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First Part: Data Preprocessing

plink –file Qatari156_filtered_pruned –maf 0.05 –geno 0.001 –hwe 0.00001 –make-bed –out cleaned_data 12509 variants removed due to missing genotype data (–geno). hwe: 0 variants removed due to Hardy-Weinberg exact test. 0 variants removed due to minor allele threshold(s)

Figure 1: Caption for the image

./plink –bfile cleaned_data –indep-pairwise 100 5 0.1 –out pruned_data

- Trying different Window Size (First Parameter) — —
- At window size = 100

```
Pruned 220 variants from chromosome 21, leaving 669.

Pruned 224 variants from chromosome 22, leaving 702.

Pruned 402 variants from chromosome 23, leaving 950.

Pruning complete. 15054 of 55226 variants removed.

Marker lists written to pruned_data.prune.in and pruned_data.prune.out .
```

Figure 2: Caption for the image

15054 of 55226 variants removed

Figure 3: Caption for the image

Figure 4: Caption for the image

```
Pruned 228 variants from chromosome 19, leaving 807.
Pruned 450 variants from chromosome 20, leaving 1115.
Pruned 232 variants from chromosome 21, leaving 657.
Pruned 241 variants from chromosome 21, leaving 685.
Pruned 461 variants from chromosome 23, leaving 891.
Pruning complete. 16031 of 55226 variants removed.

Activate Windows
Marker Lists written to pruned data prune; in and pruned data, prune.out .
```

Figure 5: Caption for the image

• At window size = 200

16031 Variants were removed

• At window size = 50

14232 of 55226 variants removed

```
www.cog-genomics.org/plink/1.9/
GNU General Public Liess
PS C:\Program Files (x86)\PLINK> ./plink --bfil
PLINK v1.90b7.2 64-bit (11 Dec 2023)
(C) 2005-2023 Shaun Purcell, Christopher Chang
                                                                               -bfile cleaned_data
 Logging to pruned_data.log.
 Options in effect:
    --bfile cleaned_data
     --indep-pairwise 50 5 0.1
    --out pruned_data
 16304 MB RAM detected; reserving 8152 MB for main workspace.
16304 MB RAM detected; reserving 8152 MB for main workspace.
Allocated 6114 MB successfully, after larger attempt(s) failed.
55226 variants loaded from .bim file.
156 people (49 males, 107 females) loaded from .fam.
Using 1 thread (no multithreaded calculations invoked).
Before main variant filters, 156 founders and 0 nonfounders present.
Calculating allele frequencies... done.
Warning: 1118 het. haploid genotypes present (see pruned_data.hh); many commands treat these as missing.
Total genotyping rate is exactly 1.
55226 variants and 156 people pass filters and 0C.
 55226 variants and 156 people pass filters and QC.
 Note: No phenotypes present.
Pruned 1081 variants from chromosome 1, leaving 3141.
Pruned 1157 variants from chromosome 2, leaving 3023.
Pruned 952 variants from chromosome 3, leaving 2622.
Pruned 884 variants from chromosome 4, leaving 2457.
 Pruned 893 variants from chromosome 5,
Pruned 853 variants from chromosome 6,
                                                                         leaving 2563.
                                                                         leaving 2380.
 Pruned 707 variants from chromosome 7,
Pruned 749 variants from chromosome 8,
                                                                         leaving 2118.
                                                                         leaving 1991.
 Pruned 613 variants from chromosome 9,
Pruned 796 variants from chromosome 10,
                                                                        leaving 1865.
                                                                          leaving 2103.
Pruned 796 variants from chromosome 10,
Pruned 688 variants from chromosome 12,
Pruned 527 variants from chromosome 13,
Pruned 509 variants from chromosome 14,
                                                                          leaving 1900.
                                                                          leaving 2096.
                                                                           leaving 1545.
                                                                          leaving 1402.
 Pruned 454 variants from chromosome 15,
                                                                           leaving
                                                                                         1356.
 Pruned 487 variants from chromosome 16,
                                                                           leaving 1422.
 Pruned 364 variants from chromosome 17,
                                                                           leaving 1274.
 Pruned 455 variants from chromosome 18,
                                                                           leaving 1349.
 Pruned 205 variants from chromosome 19,
                                                                          leaving 830.
 Pruned 392 variants from chromosome 20,
Pruned 213 variants from chromosome 21,
                                                                          leaving 1173.
                                                                          leaving 676.
 Pruned 211 variants from chromosome 22,
                                                                          leaving
                                                                                         715.
 Pruned 359 variants from chromosome 23, leaving 993
Pruning complete. 14232 of 55226 variants removed.
                                                                          leaving 993.
 Marker lists written to pruned_data.prune.in and pruned_data.prune.out
```

Figure 6: Caption for the image

- Trying different Step Size (Second Parameter) —————
- At step size = 5

15054 of 55226 variants removed

• At step size = 20

```
Pruned 220 variants from chromosome 21, leaving 669.
Pruned 224 variants from chromosome 22, leaving 702.
Pruned 402 variants from chromosome 23, leaving 950.
Pruning complete. 15054 of 55226 variants removed.
Marker lists written to pruned_data.prune.in and pruned_data.prune.out .

DS C:\Program Files (x86)\DITMY>
```

Figure 7: Caption for the image

Figure 8: Caption for the image

```
./plink --bfile cleaned_data --indep-pairwise 100 20 0.1 --out pruned_data
PLINK v1.90b7.2 64-bit (11 Dec 2023) www.cog-genomics.org/plink/1.9/
(C) 2005-2023 Shaun Purcell, Christopher Chang GNU General Public License v3
Logging to pruned_data.log.
Options in effect:
--bfile cleaned_data.
                                     --bfile cleaned_data
--indep-pairwise 100 20 0.1
--out pruned_data
--indep-pairwise 100 20 0.1
--out pruned_data

16304 MB RAM detected; reserving 8152 MB for main workspace.
Allocated 6114 MB successfully, after larger attempt(s) failed.
55226 variants loaded from .bim file.
156 people (49 males, 107 females) loaded from .fam.
Using 1 thread (no multithreaded calculations invoked).
Before main variant filters, 156 founders and 0 nonfounders present.
Calculating allele frequencies... done.
Warning: 1118 het. haploid genotypes present (see pruned_data.hh ); many commands treat these as missing.
Total genotyping rate is exactly 1.
55226 variants and 156 people pass filters and QC.
Note: No phenotypes present.
Pruned 1135 variants from chromosome 1, leaving 3087.
Pruned 1907 variants from chromosome 2, leaving 2968.
Pruned 1912 variants from chromosome 3, leaving 2567.
Pruned 921 variants from chromosome 4, leaving 2567.
Pruned 991 variants from chromosome 6, leaving 2594.
Pruned 992 variants from chromosome 6, leaving 2333.
Pruned 728 variants from chromosome 7, leaving 2097.
Pruned 784 variants from chromosome 7, leaving 1956.
Pruned 785 variants from chromosome 10, leaving 1869.
Pruned 786 variants from chromosome 11, leaving 1869.
Pruned 730 variants from chromosome 11, leaving 1869.
Pruned 731 variants from chromosome 12, leaving 1869.
Pruned 732 variants from chromosome 13, leaving 1512.
Pruned 560 variants from chromosome 14, leaving 1373.
Pruned 571 variants from chromosome 15, leaving 1373.
Pruned 372 variants from chromosome 16, leaving 1397.
Pruned 373 variants from chromosome 17, leaving 1399.
Pruned 374 variants from chromosome 17, leaving 1690.
Pruned 375 variants from chromosome 17, leaving 1690.
Pruned 387 variants from chromosome 19, leaving 1397.
Pruned 210 variants from chromosome 19, leaving 1397.
Pruned 211 variants from chromosome 19, leaving 1399.
Pruned 220 variants from chromosome 21, leaving 1690.
Pruned 220 variants from chromosome 22, leaving 963.
Pruned 230 variants from chromosome 23, leaving 963.
Pruned 230 variants from chromosome 24, leaving 963.
```

14958 of 55226 variants removed.

• At step size = 1

Pruning complete. 15066 of 55226 variants removed.

- Trying different LD (Third Parameter) —————
- LD = .1
- LD = .9

Figure 9: Caption for the image

```
Pruned 220 variants from chromosome 21, leaving 669.

Pruned 224 variants from chromosome 22, leaving 702.

Pruned 402 variants from chromosome 23, leaving 950.

Pruning complete. 15054 of 55226 variants removed.

Marker lists written to pruned_data.prune.in and pruned_data.prune.out .
```

Figure 10: Caption for the image

```
PS C:\Program Files (x86)\PLINK> ./plink --bfile cleaned.data --indep-pairwise 180 5 8.1 --out pruned.data
PLINK v1.98b7.2 64-bit (11 Dec 2823)
    www.cog-genomics.org/plink/1.9/
(C) 2805-2823 Shaun Purcell, Christopher Chang GNU General Public License v3
Logging to pruned_data.log.
Options in effect:
    --bfile cleaned_data
    --indep-pairwise 180 5 0.1
    --out pruned_data
    --out pruned_da
```

Figure 11: Caption for the image

0 of 55226 variants removed.

• LD = .01

53685 of 55226 variants removed.

So when we: ////* Increase/Decrease window size -> more/less SNPs are removed Increase/Decrease step size -> less/more SNPs are removed Increase/Decrease LD -> less/more SNPs are removed

• We'll do the pca on the result of window size = 100, step size = 5, LD = .1 We had to recode the

Figure 12: Caption for the image

result of this operation to be able to find the data in ped and map format: ./plink -bfile cleaned_data -indep-pairwise 100 5 0.1 -out pruned_data -recode This is the result of the operation

Now we do this operation to make sure we got the right SNPs

PART II: Identify SNPs associated with population structure

I used this video in this part for guidance:

https://www.youtube.com/watch?v=vos6VeuNcaM&ab_channel=GenomicsBootCamp

• First we run PCA on the cleaned_data ./plink -bfile cleaned_data -pca -out pca_results

This is the result:

• Second we read the .raw file C:/Program Files (x86)/PLINK/

library(tidyverse)

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
        1.1.4 v readr
## v dplyr
                             2.1.5
## v forcats 1.0.0
                   v stringr
                             1.5.1
## v ggplot2 3.4.4
                   v tibble
                             3.2.1
## v lubridate 1.9.3
                   v tidyr
                             1.3.1
           1.0.2
## v purrr
## -- Conflicts ------ tidyverse_conflicts() --
```

```
PS C:\Program Files (x86)\PLINK> ./plink --file pruned_data --pca --out pca_results
PLINK v1.90b7.2 64-bit (11 Dec 2023)
                                                      www.cog-genomics.org/plink/1.9/
(C) 2005-2023 Shaun Purcell, Christopher Chang
                                                        GNU General Public License v3
Logging to pca_results.log.
Options in effect:
  --file pruned_data
  --out pca_results
  --рса
16304 MB RAM detected; reserving 8152 MB for main workspace.
.ped scan complete (for binary autoconversion).
Performing single-pass .bed write (55226 variants, 156 people).
--file: pca_results-temporary.bed + pca_results-temporary.bim +
pca_results-temporary.fam written.
55226 variants loaded from .bim file.
156 people (49 males, 107 females) loaded from .fam.
Using up to 11 threads (change this with --threads).
Before main variant filters, 156 founders and 0 nonfounders present. Calculating allele frequencies... done.
Warning: 1118 het. haploid genotypes present (see pca_results.hh ); many
commands treat these as missing.
Total genotyping rate is exactly 1. 55226 variants and 156 people pass filters and QC.
Note: No phenotypes present.
Excluding 1352 variants on non-autosomes from relationship matrix calc.
Relationship matrix calculation complete.
--pca: Results saved to pca_results.eigenval and pca_results.eigenvec
```

Figure 13: Caption for the image

Figure 14: Caption for the image

Figure 15: Caption for the image

```
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                      masks stats::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
library(caret)
## Warning: package 'caret' was built under R version 4.3.3
## Loading required package: lattice
##
## Attaching package: 'caret'
##
## The following object is masked from 'package:purrr':
##
##
       lift
library(scatterplot3d)
#install.packages("qqman")
library(qqman)
## Warning: package 'qqman' was built under R version 4.3.3
##
## For example usage please run: vignette('qqman')
##
## Citation appreciated but not required:
## Turner, (2018). qqman: an R package for visualizing GWAS results using Q-Q and manhattan plots. Jour.
```

```
##
##
## Attaching package: 'qqman'
## The following object is masked from 'package:lattice':
##
       qq
genotype_data <- read.table("C:/Program Files (x86)/PLINK/recoded_data.raw", header = TRUE, sep = "")</pre>
genotype_data_filtered <- genotype_data[, c(1,7:ncol(genotype_data))]</pre>
t <- read.table("C:/Program Files (x86)/PLINK/recoded_data.raw", header = TRUE, sep = "")
eigval <- read.table("C:/Program Files (x86)/PLINK/pca results.eigenval", header = FALSE, sep = "")
eigvec <- read.table("C:/Program Files (x86)/PLINK/pca_results.eigenvec", header = FALSE, sep = "")
df <- data.frame(x = genotype_data_filtered[, c(2: ncol(genotype_data_filtered))], y = eigvec[c(3: ncol</pre>
I'vw commented the following parts to be able to knit the file, as they may take 4~5 hours to finish
#num_tests <- (ncol(genotype_data_filtered) - 1) * 3</pre>
\#result\_df \leftarrow data.frame(SNP = character(num\_tests), P = numeric(num\_tests), PC = integer(num\_tests), B
#ptr = 1
#for (i in 3:5) {
  #for (j in 2:ncol(genotype_data_filtered)) {
    #for (k in 3:5) {
      #for (l in 3:5) {
        #if (i != k & i != l & k != l & k < l) {
          # Construct formula
        \# formula \leftarrow reformulate(c(paste0("x.", names(genotype_data_filtered)[j]), paste0("y.V", k), p
          # Fit linear regression model
         # model <- lm(formula, data = df)</pre>
          # Extract p-value, beta coefficient, and standard error
        # summary_coef <- summary(model)$coefficients</pre>
          \#p\_value \leftarrow summary\_coef[2, "Pr(>|t|)"]
          # beta <- summary_coef[2, "Estimate"]</pre>
        # se <- summary_coef[2, "Std. Error"]</pre>
           # Store results in result_df
                      result\_df[ptr, ] \leftarrow c(names(genotype\_data\_filtered)[j], p\_value, i - 2, beta, se)
           ptr = ptr + 1
           print(ptr)
         }
   #
      }
```

```
#}
#}
# Print the first few rows of the result data frame
#print(head(result_df))
map data <- read.table("C:/Program Files (x86)/PLINK/map.map", header = FALSE, sep = "\t", col.names =</pre>
\#modified\_result\_df \leftarrow result\_df
\#modified\_result\_df\$SNP \leftarrow substring(result\_df\$SNP, 1, nchar(result\_df\$SNP) - 2)
#merqed_data <- merqe(modified_result_df, map_data, by.x = "SNP", by.y = "SNP_ID", all.x = TRUE)</pre>
#merged_data$P <- as.numeric(merged_data$P)</pre>
#str(merged_data)
#pc1_data <- merged_data[merged_data$PC == 1, ]</pre>
#pc2_data <- merged_data[merged_data$PC == 2, ]</pre>
#pc3_data <- merged_data[merged_data$PC == 3, ]</pre>
\#manhattan\_data\_1 \leftarrow pc1\_data[, c("SNP", "Chromosome", "Physical\_Position", "P")]
\#colnames(manhattan\_data\_1) \leftarrow c("SNP", "CHR", "BP", "P")
#manhattan_data_1$P <- as.numeric(manhattan_data_1$P)</pre>
# Create the QQMan Manhattan plot
\#manhattan(manhattan_data_1)
```

This was the result:

Now we'll sort the PC1 dataframe

```
#pc1_data_sorted <- pc1_data[order(pc1_data$P), ]
#pc1_data_top10 <- pc1_data_sorted[1:10, ]
#pc1_data_top10
#write.csv(pc1_data_top10, "j:/output_file.csv", row.names = TRUE)</pre>
```

This was the result

Now after searching in the dbSNP, using excel we constructed this table

Now we'll study the SNPs with the second PC, but for the next we only will draw the manhattan and show the top 10

This was the result:

Again, we'llsort the resuls and get the top 10 SNPs and find the data about them

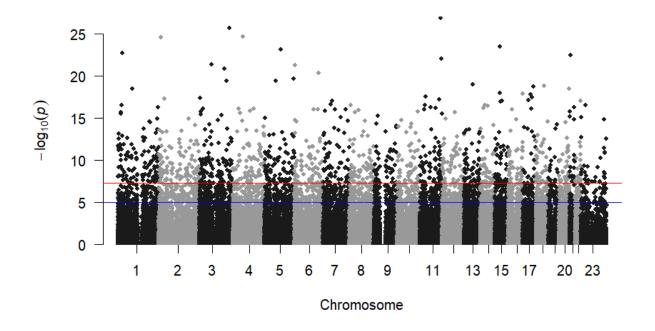


Figure 16: Caption for the image

l_data_to l_data_to	pp10 <- pc1_data_so pp10	rder(pc1_data\$P),] rted[1:10,] /output_file.csv", ı	row.n	ames = TRUE)				
Description:	df [10 × 8]							@ & :
	SNP <chr></chr>	P <dbl></dbl>	PC <chr></chr>	Beta <chr></chr>	SE <chr></chr>	Chromosome	Genetic_Distance	Physical_Position
5772	rs10466604	1.509941e-27	1	0.105403113948956	0.0078616268882563	11	124.15914	124159136
139762	rs7355960	2.450214e-26	1	0.133630459485203	0.0103135952990324	3	180.56674	180566740
94081	rs335339	2.612128e-25	1	0.153248688973709	0.0121869103318538	4	62.01347	62013467
49040	rs16857866	2.795157e-25	1	0.141137661458271	0.0112335519065755	2	11.82817	11828169
68491	rs1841575	3.659841e-24	1	0.149129102873668	0.012274693195495	15	51.88696	51886958
91032	rs291799	7.881589e-24	1	0.165295497232791	0.0137452190084827	5	96.76758	96767581
21316	rs11247683	2.243430e-23	1	0.11219697781958	0.00946247578592448	1	27.97429	27974294
88861	rs2825326	3.921665e-23	1	0.15239591047397	0.0129511430281785	21	19.34445	19344445
134997	rs7129025	9.111972e-23	1	0.150848303288673	0.0129695049743154	11	131.07117	131071174
41357	rs1388277	5.190549e-22	1	0.173947283349466	0.0153255850538021	3	73.20699	73206990

Figure 17: co

SNP	~ P	~	Beta 🔻 🤄	SE 🔻	Chromosome 🔻 (Genetic_Distance 🔻	Physical_Position Gene	ALFA Freq	Min Freq	Max Freq 🔻
rs10466604		1.51E-27	0.105403114	0.007861627	11	124.15914	124159136 MSANTD2	0.255752/4958	0.171667/103 (NorthernSweden) G=0.416667/5 (Siberian)
rs7355960		2.45E-26	0.133630459	0.010313595	3	180.56674	180566740 MFN1	T=0.019974/5592	T=0.001563/7 (Estonian)	T=0.317708/366 (HapMap)
rs335339		2.61E-25	0.153248689	0.01218691	4	62.013467	62013467 ADGRL3	G=0.038261/1759 (ALFA)	G=0./0 (GENOME_DK)	G=0.215136/253 (HapMap)
rs16857866		2.80E-25	0.141137661	0.011233552	2	11.828169	11828169 LPIN1	T=0.077761/4340 (ALFA)	T=0.000342/1 (KOREAN)	C=0.322222/29 (SGDP_PRJ)
rs1841575		3.66E-24	0.149129103	0.012274693	15	51.886958	51886958 none	C=0.071773/18046 (ALFA	C=0.017857/1 (Siberian)	C=0.222751/421 (HapMap)
rs291799		7.88E-24	0.165295497	0.013745219	5	96.767581	96767581 none	A=0.054396/1178 (ALFA)	A=0.004008/4 (GoNL)	A=0.135603/243 (HapMap)
rs11247683		2.24E-23	0.112196978	0.009462476	1	27.974294	27974294 STX12	G=0.071484/3878 (ALFA)	G=0.007143/32 (Estonian)	A=0.355072/49 (SGDP_PRJ)
rs2825326		3.92E-23	0.15239591	0.012951143	21	19.344445	19344445 none	T=0.079939/2453 (ALFA)	T=0.000223/1 (Estonian)	C=0.4375/21 (SGDP_PRJ)
rs7129025		9.11E-23	0.150848303	0.012969505	11	131.07117	131071174 NTM	T=0.01692/4577 (ALFA)	T=0./0 (PRJEB36033)	T=0./0 (PRJEB36033)
rs1388277		5.19E-22	0.173947283	0.015325585	3	73.20699	73206990 none	G=0.026358/1280 (ALFA)	G=0.000259/1 (ALSPAC)	A=0.5/12 (SGDP_PRJ)

Figure 18: c

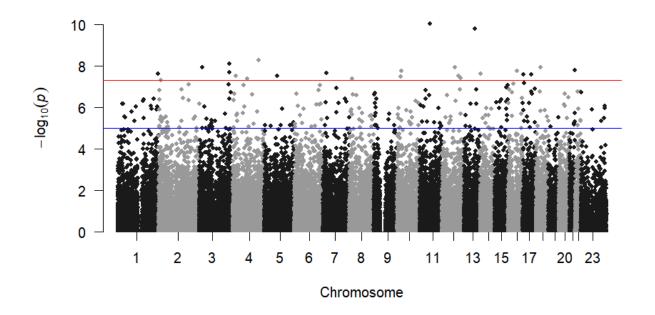


Figure 19: Caption for the image

```
#pc2_data_sorted <- pc2_data[order(pc2_data$P), ]
#pc2_data_top10 <- pc2_data_sorted[1:10, ]
#pc2_data_top10
#write.csv(pc2_data_top10, "j:/pc2_file.csv", row.names = TRUE)</pre>
```

{r} 2_data_sc 2_data_tc 2_data_tc	pp10 <- pc2_data_so	order(pc2_data\$P),] orted[1:10,]					⊕ ¥)
Description:	df [10 × 8]						@
	SNP <chr></chr>	P PC <dbl> <chr></chr></dbl>	Beta <chr></chr>	SE <chr></chr>	Chromosome	Genetic_Distance	Physical_Position
97465	rs3815045	8.946506e-11 2	0.100737842839959	0.0144492975691902	11	61.42652	61426522
105416	rs4536348	1.613955e-10 2	0.120083386201646	0.0175001459677472	13	80.46093	80460925
90366	rs2880416	5.410293e-09 2	0.075271960181628	0.012165121552094	4	156.34399	156343993
90203	rs28711160	7.653785e-09 2	0.0504125013776107	0.00823893673669989	3	178.91682	178916822
34045	rs12637343	1.180414e-08 2	0.0735074395232198	0.0121854652949839	3	15.36482	15364820
138911	rs7308149	1.206585e-08 2	0.0612672111174184	0.0101637812828436	12	76.21468	76214682
102148	rs4258695	1.219666e-08 2	0.096866116537736	0.0160751410390226	18	29.88665	29886646
152603	rs8131179	1.574500e-08 2	-0.0451291701798742	0.00755358953960826	21	42.95527	42955270
157682	rs936873	1.698660e-08 2	-0.0819036206572144	0.0137439789294814	16	53.70562	53705623
38685	rs1326644	1.741805e-08 2	-0.0482604980048717	0.00810531839297284	10	24.43515	24435153

Figure 20: co

SNP	₩ P	~	Beta 🔻	SE	Chromosome 💌	Genetic_Distance 💌	Physical_Position Gene	ALFA Freq	Min Freq	Max Freq 🔻
rs3815045		8.95E-11	0.100737843	0.014449298	11	61.426522	61426522 RAB3IL1	A=0.039392/14813 (ALFA)	A=0./0 (PRJEB36033)	G=0.45/9 (Siberian)
rs4536348		1.61E-10	0.120083386	0.017500146	13	80.460925	80460925 none	C=0.143399/29940 (ALFA)	C=0.050926/11 (Qatari)	T=0.5/5 (Siberian)
rs2880416		5.41E-09	0.07527196	0.012165122	4	156.34399	156343993 NPY2R (Varview), NPY2R-AS1 (Varview)) G=0.200635/3790	G=0.143519/31 (Qatari)	G=0.454243/7613 (TOMMO)
rs28711160		7.65E-09	0.050412501	0.008238937	3	178.91682	178916822 LINC00578	G=0.413223/10619 (ALFA)	A=0.212264/45 (Vietnamese)	A=0.475/19 (GENOME_DK)
rs12637343		1.18E-08	0.07350744	0.012185465	3	15.36482	15364820 none	C=0.138514/13314 (ALFA)	C=0.078704/17 (Qatari)	T=0.428571/12 (Siberian)
rs7308149		1.21E-08	0.061267211	0.010163781	12	76.214682	76214682 none	G=0.255161/4820 (ALFA)	G=0.166667/36 (Qatari)	G=0.466983/7827 (TOMMO)
rs4258695		1.22E-08	0.096866117	0.016075141	18	29.886646	29886646 NOL4	G=0.047383/1684 (ALFA)	G=0.033333/20 (NorthernSweden) T=0.5/3 (Siberian)
rs8131179		1.57E-08	-0.04512917	0.00755359	21	42.95527	42955270 PDE9A	T=0.359551/22285 (ALFA)	T=0.188498/118 (Chileans)	C=0.469809/887 (HapMap)
rs936873		1.70E-08	-0.081903621	0.013743979	16	53.705623	53705623 none	C=0.068097/10342 (ALFA)	C=0.028219/32 (Daghestan)	C=0.197555/307 (HapMap)
rs1326644		1.74E-08	-0.048260498	0.008105318	10	24.435153	24435153 KIAA1217	A=0.137643/28731	A=0.044377/744 (TOMMO)	G=0.5/5 (Siberian)

Figure 21: co

Here is the resulting one:

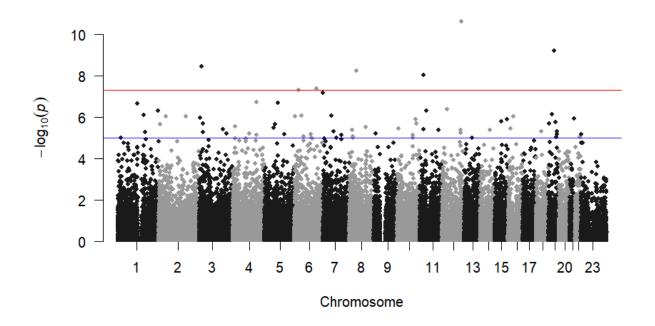


Figure 22: co

```
#pc3_data_sorted <- pc3_data[order(pc3_data$P), ]
#pc3_data_top10 <- pc3_data_sorted[1:10, ]
#pc3_data_top10</pre>
```

Here is the resulting table:

It's enought to search for the previous 2 tables in dbSNP.

• Task 2.2 ---*-**- We know from our previous knowledge about PCA that the new coordinates of the points are aquired by: D' = UtDt, where t is for transpose We know also that the U matrix is a d*r matrix. where d is the number of dimensions in the original space, and r is the new required dimensions of the reduced space. We want to map this to our problem, r should be 3 here, and if the PCA function of PLINK returns U not D' we should have a matrix of 20 columns, and number of rows equal to number of SNPs which is ~56000. However, we see that the eigvec is consisting of 156 row, which is the number of samples, or n.

	pp10 <- pc3_data_sc	order(pc3_data\$P),] orted[1:10,]					⊕ ¥
Description:	df [10 × 8]						<i>a</i> *
	SNP <chr></chr>	P PC <dbl> <chr></chr></dbl>	Beta <chr></chr>	SE <chr></chr>	Chromosome	Genetic_Distance	Physical_Position
12458	rs10850824	2.382363e-11 3	0.0575301024202837	0.00797313262574598	12	116.349440	116349444
52598	rs16963743	6.435932e-10 3	0.0954586819602495	0.0144622087700657	19	35.259505	35259505
107667	rs4684859	3.532565e-09 3	0.0532499505950568	0.00849093963543066	3	12.473401	1247340
75766	rs2122950	5.649665e-09 3	0.0891173142581944	0.0144226805752142	8	40.601978	40601978
25716	rs11820583	9.216243e-09 3	0.0564599228958268	0.00928336342668996	11	21.277586	21277586
77729	rs2204732	4.298353e-08 3	-0.0474580980013374	0.00822460034363213	6	131.502620	13150262
87530	rs2744278	4.966115e-08 3	0.0895672330577778	0.0156025702515097	6	25.391012	2539101
146375	rs7793347	6.549559e-08 3	0.0649545515945411	0.0114289584312027	7	0.270131	27013
61056	rs17323440	1.904495e-07 3	0.0543606620166025	0.0099573655957616	4	141.253340	14125334
160885	rs958535	2.021182e-07 3	0.102047009154674	0.0187355687569845	5	82.639636	8263963

Figure 23: co

```
Xr <- t(eigvec[, 3:5])
Y <- t(genotype_data_filtered[, 2:ncol(genotype_data_filtered)])
nrow(Xr)

## [1] 3

ncol(Xr)

## [1] 156

nrow(Y)

## [1] 55226

ncol(Y)</pre>
```

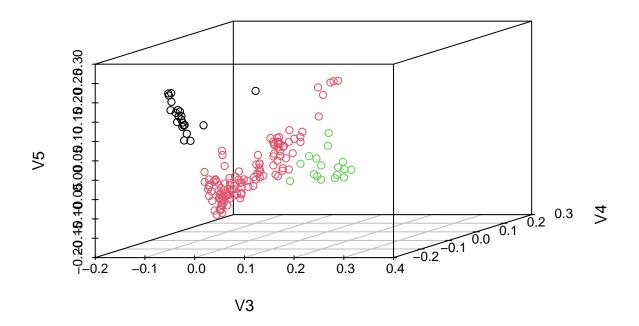
We deduce that PLINK returns the matrix D', dimensions of points in the new space of 3d. So we should apply the clustering on it directly. I asked chatgpt to make sure of this, it confirmed. PLINK's `--pca` option returns the principal component (PC) scores for each individual in the dataset, which represent the coordinates of each individual in the reduced dimensional space defined by the principal components. It does not directly return the eigenvectors or eigenvalues.

The output typically includes:

[1] 156

```
set.seed(123)
k <- 3
kmeans_result <- kmeans(eigvec[, 3:5], centers = k)
scatterplot3d(eigvec[, 3:5], color = kmeans_result$cluster, main = "3D Scatterplot with K-Means Cluster")</pre>
```

3D Scatterplot with K-Means Clustering



```
new_df <- as.data.frame(eigvec[, 3:5])

# Create one-hot encoded cluster labels
cluster_labels <- matrix(0, nrow = nrow(new_df), ncol = k)
for (i in 1:k) {
    cluster_labels[kmeans_result$cluster == i, i] <- 1
}

# Add one-hot encoded cluster labels to the new dataframe
colnames(cluster_labels) <- pasteO("cluster", 1:k)
new_df <- cbind(new_df, cluster_labels)</pre>
```

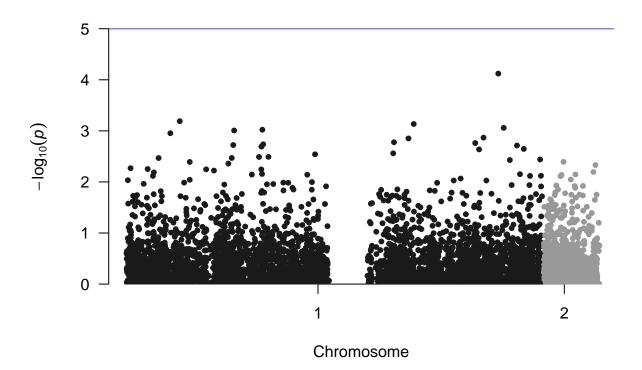
For saving time, the TA permitted us to do the work on 1000 - 10000 SNP instead of all of the data set, we'll work on 5000.

```
selected_columns <- genotype_data_filtered[, seq(2, ncol(genotype_data_filtered), by = 11)]
data <- cbind(new_df, selected_columns)</pre>
```

```
num_tests <- (ncol(data)) * 3
result_df1 <- data.frame(SNP = character(num_tests), P = numeric(num_tests), cluster = integer(num_test
ptr = 1

for (i in 1:k) {
   for (j in 7:ncol(data)) {
     formula <- reformulate(c(paste0(names(data)[j]), paste0("V", 3), paste0("V", 4), paste0("V", 5)), r</pre>
```

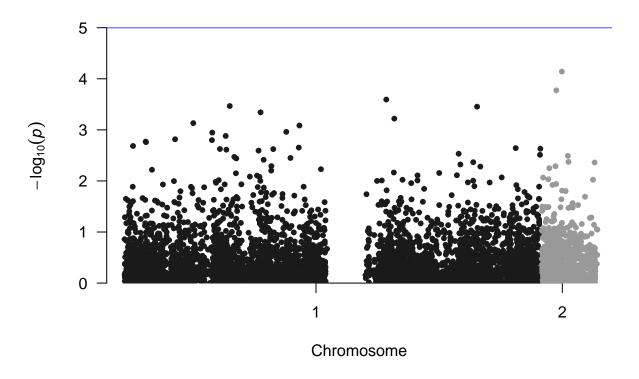
```
model <- lm(formula, data = data)</pre>
    summary_coef <- summary(model)$coefficients</pre>
    p_value <- summary_coef[2, "Pr(>|t|)"]
    beta <- summary_coef[2, "Estimate"]</pre>
    se <- summary_coef[2, "Std. Error"]</pre>
    result_df1[ptr, ] <- c(names(genotype_data_filtered)[j], p_value, i, beta, se)</pre>
    ptr = ptr + 1
}
modified_result_df1 <- result_df1</pre>
modified_result_df1$SNP <- substring(result_df1$SNP, 1, nchar(result_df1$SNP) - 2)</pre>
merged_data1 <- merge(modified_result_df1, map_data, by.x = "SNP", by.y = "SNP_ID", all.x = TRUE)</pre>
merged_data1 <- na.omit(merged_data1)</pre>
merged_data1$P <- as.numeric(merged_data1$P)</pre>
c1_data <- merged_data1[merged_data1$cluster == "1", ]</pre>
c2_data <- merged_data1[merged_data1$cluster == "2", ]</pre>
c3_data <- merged_data1[merged_data1$cluster == "3", ]</pre>
manhattan_data_1 <- c1_data[, c("SNP", "Chromosome", "Physical_Position", "P")]</pre>
colnames(manhattan_data_1) <- c("SNP", "CHR", "BP", "P")</pre>
manhattan_data_1$P <- as.numeric(manhattan_data_1$P)</pre>
# Create the QQMan Manhattan plot
manhattan(manhattan_data_1)
```



```
c1_data_sorted <- c1_data[order(c1_data$P), ]
c1_data_top10 <- c1_data_sorted[1:10, ]
c1_data_top10</pre>
```

```
P cluster
##
                SNP
                                                                             SE
                                                        Beta
## 6122 rs17162892 7.560011e-05
                                       1 0.0866962622210513 0.0213019793355854
## 14282 rs804429 6.475575e-04
                                       1 -0.0621534366554175 0.0178417985831305
## 5341 rs16844658 7.340055e-04
                                       1 -0.0707296246838893 0.0205180832739289
## 9572 rs3913657 8.727141e-04
                                       1 0.0597698425679151 0.0175978156668449
                                       1 -0.0467057475360225 0.0138508099293416
## 12244 rs6671683 9.477386e-04
## 11299
         rs555146 9.836841e-04
                                       1 0.0487687651396717 0.0145101371826581
                                           0.051925594743484 0.0156189330744395
## 10858 rs4908343 1.111680e-03
## 5743 rs17020918 1.355751e-03
                                       1 -0.0517753429763544 0.0158591723478737
## 14923
         rs961404 1.406868e-03
                                       1 -0.0641849722296378 0.0197282958483413
                                       1 -0.038814668408714 0.012126850922408
## 2578 rs11587040 1.671335e-03
##
         Chromosome Genetic_Distance Physical_Position
## 6122
                 1
                          220.42402
                                             220424015
## 14282
                           33.33810
                 1
                                              33338100
## 5341
                 1
                           170.74873
                                             170748734
## 9572
                 1
                           223.59646
                                             223596463
## 12244
                 1
                            81.84146
                                              81841463
## 11299
                  1
                            65.29999
                                              65299995
## 10858
                 1
                           27.80429
                                              27804285
                 1
## 5743
                          211.77346
                                             211773462
## 14923
                  1
                           167.67247
                                             167672474
## 2578
                           159.16641
                                             159166407
```

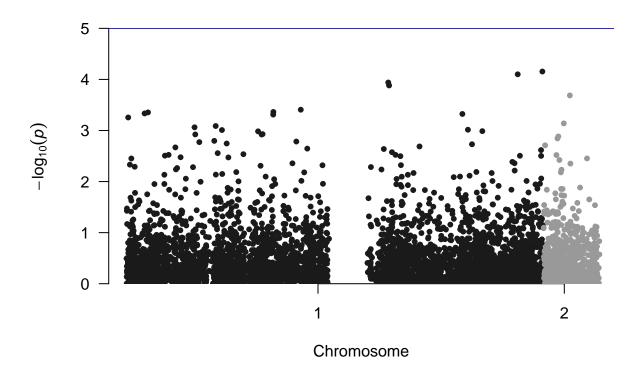
```
manhattan_data_2 <- c2_data[, c("SNP", "Chromosome", "Physical_Position", "P")]
colnames(manhattan_data_2) <- c("SNP", "CHR", "BP", "P")
manhattan_data_2$P <- as.numeric(manhattan_data_2$P)
manhattan(manhattan_data_2)</pre>
```



No one again is significantly associated with second cluster

```
c2_data_sorted <- c2_data[order(c2_data$P), ]
c2_data_top10 <- c2_data_sorted[1:10, ]
c2_data_top10</pre>
```

```
SNP
                                                                               SE
##
                                P cluster
                                                          Beta
## 5498
         rs16857866 7.227758e-05
                                        2
                                           -0.123323034152977
                                                               0.030215251367673
## 9157
           rs345903 1.685911e-04
                                           0.0769376203809053 0.0199398773382473
                                        2
## 7701
          rs2244798 2.548312e-04
                                           -0.178407619417091
                                                               0.047622427672023
## 7748
          rs2269252 3.413882e-04
                                        2
                                            0.105855789863394 0.0288797511105179
  14651
           rs924569 3.511197e-04
                                        2 -0.0717764529648769 0.0196241648420027
                                            0.117295293031882 0.0327143030517655
  3377 rs12090448 4.537164e-04
                                        2 - 0.0694915618341067 \ 0.0198303133491292
  13516
         rs7525955 6.027103e-04
  12133
          rs6665972 7.377420e-04
                                          0.0721657692523048 0.0209436945324718
## 2932
        rs12021686 8.228888e-04
                                        2 -0.0915899554837591 0.0268298510324016
## 3209
         rs12060538 1.094881e-03
                                        2 0.0785154506940166 0.0235846841048672
         Chromosome Genetic Distance Physical Position
##
## 5498
                  2
                            11.82817
                                               11828169
                  2
## 9157
                             8.60861
                                                8608610
## 7701
                  1
                           155.79459
                                              155794587
## 7748
                  1
                            63.87065
                                               63870653
## 14651
                  1
                           209.14624
                                              209146241
## 3377
                  1
                            81.97615
                                               81976150
## 13516
                  1
                           160.51446
                                              160514456
## 12133
                            42.44611
                                               42446110
```



One only SNP is significantly associated to the third cluster we shall search for it.

```
c3_data_sorted <- c3_data[order(c3_data$P), ]
c3_data_top10 <- c3_data_sorted[1:10, ]
c3_data_top10</pre>
```

```
##
                SNP
                               P cluster
                                                        Beta
                                                                             SE
## 11879 rs6587420 7.001009e-05
                                            0.12764665504888 0.0312116970971629
          rs701228 7.936884e-05
                                       3
                                           -0.10931198581659 0.0269422442730761
## 13110
## 7699
         rs2244798 1.148761e-04
                                       3
                                           0.140508802695761 0.0354766831124132
## 3211 rs12061312 1.312542e-04
                                       3 -0.122609365933987 0.0312354015702992
## 225
         rs10206116 2.056257e-04
                                       3
                                           0.117655094639041 0.0309223852742068
## 6571 rs17533693 3.918598e-04
                                       3 0.0974274413365295 0.0268631853975606
## 10070 rs4471236 4.334250e-04
                                       3 -0.0869579861890864 0.0241657943419837
## 2536 rs11584308 4.421008e-04
                                       3 0.0696714304688626 0.0193920290827495
## 8888
         rs3000873 4.632510e-04
                                       3 0.0500405671516608 0.0139795795819277
## 12343 rs6677721 4.741785e-04
                                           0.107104853530145 0.0299767693480647
```

##		${\tt Chromosome}$	<pre>Genetic_Distance</pre>	Physical_Position
##	11879	1	246.30915	246309151
##	13110	1	231.83561	231835613
##	7699	1	155.79459	155794587
##	3211	1	156.29316	156293163
##	225	2	15.36810	15368097
##	6571	1	104.40717	104407173
##	10070	1	88.31494	88314935
##	2536	1	14.67858	14678580
##	8888	1	12.69291	12692907
##	12343	1	199.33262	199332625

NO significant SNP in relation to any cluster. However we'll do the research in the dbSNP becquise it's required.

After researching, here is the result.



Figure 24: co

THANK YOU!