

# Wholesale Customer Segmentation Technical Document

To find out the spending pattern of the business clients, we did the below clustering analysis and built up channel and region analysis upon it.

First, we use a database that contains 440 observations from the distributor's clients.

```
## Install required packages
# install.packages('corrplot')
# install.packages('tidyverse')
```

```
library(stats)
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
##
## filter, lag
```

```
## The following objects are masked from 'package:base':
##
## intersect, setdiff, setequal, union
```

```
library(cluster)
library(ggplot2)
# library(corrplot)
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v forcats 1.0.0 v stringr 1.5.0
## v lubridate 1.9.2 v tibble 3.2.1
## v purrr 1.0.1 v tidyr 1.3.0
## v readr 2.1.4
```

```
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
# Read the data into a dataframe
wholesale = read.csv('Wholesale customers data.csv')
```

```
## Replace values for Channel
wholesale$Channel[wholesale$Channel == 1] <- 'Horeca'
```

```

wholesale$Channel[wholesale$Channel == 2] <- 'Retail'

## Replace values for Region
wholesale$Region[wholesale$Region == 1] <- 'Lisbon'
wholesale$Region[wholesale$Region == 2] <- 'Oporto'
wholesale$Region[wholesale$Region == 3] <- 'Other regions'

# Set k = the number of clusters we want
k=3

```

Performed a rudimentary exploratory data analysis on the data provided.

```

## CORRELATION ANALYSIS
corr_matrix <- cor(wholesale[,3:8])
round(corr_matrix, 2)

```

```

##           Fresh Milk Grocery Frozen Detergents_Paper Delicatessen
## Fresh           1.00 0.10   -0.01   0.35           -0.10           0.24
## Milk             0.10 1.00    0.73   0.12            0.66           0.41
## Grocery          -0.01 0.73    1.00  -0.04            0.92           0.21
## Frozen           0.35 0.12   -0.04   1.00           -0.13           0.39
## Detergents_Paper -0.10 0.66    0.92  -0.13            1.00           0.07
## Delicatessen     0.24 0.41    0.21   0.39            0.07           1.00

```

```

# Pivot data to create a single continuous/numerical factor using gather function
temp1 <- gather(data = wholesale, key = 'Product_Category', value = 'Spend', Fresh:Delicatessen)

# Summarise data for analysis
temp2 <- temp1 %>% group_by(Channel, Region, Product_Category) %>%
  summarise(TotalSpend = sum(Spend), AvgSpend = mean(Spend))

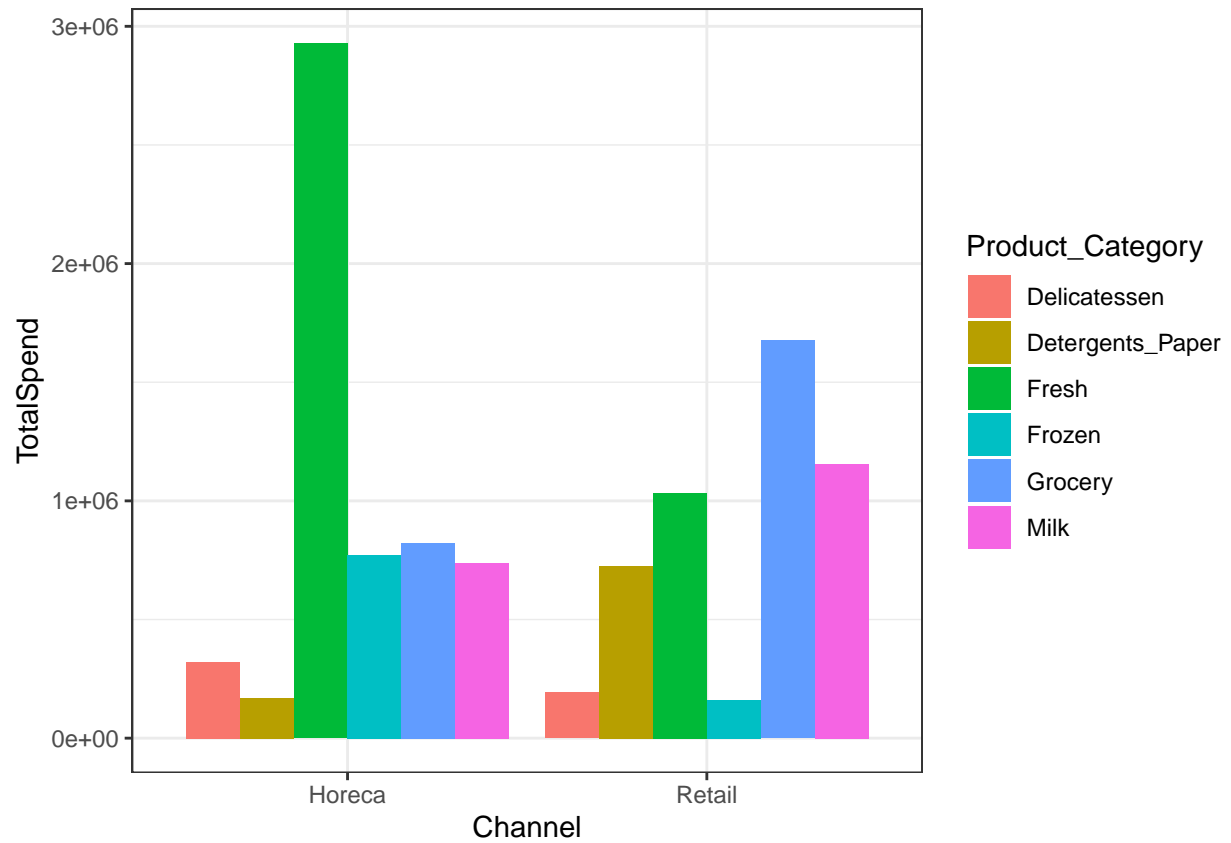
```

## 'summarise()' has grouped output by 'Channel', 'Region'. You can override using  
## the '.groups' argument.

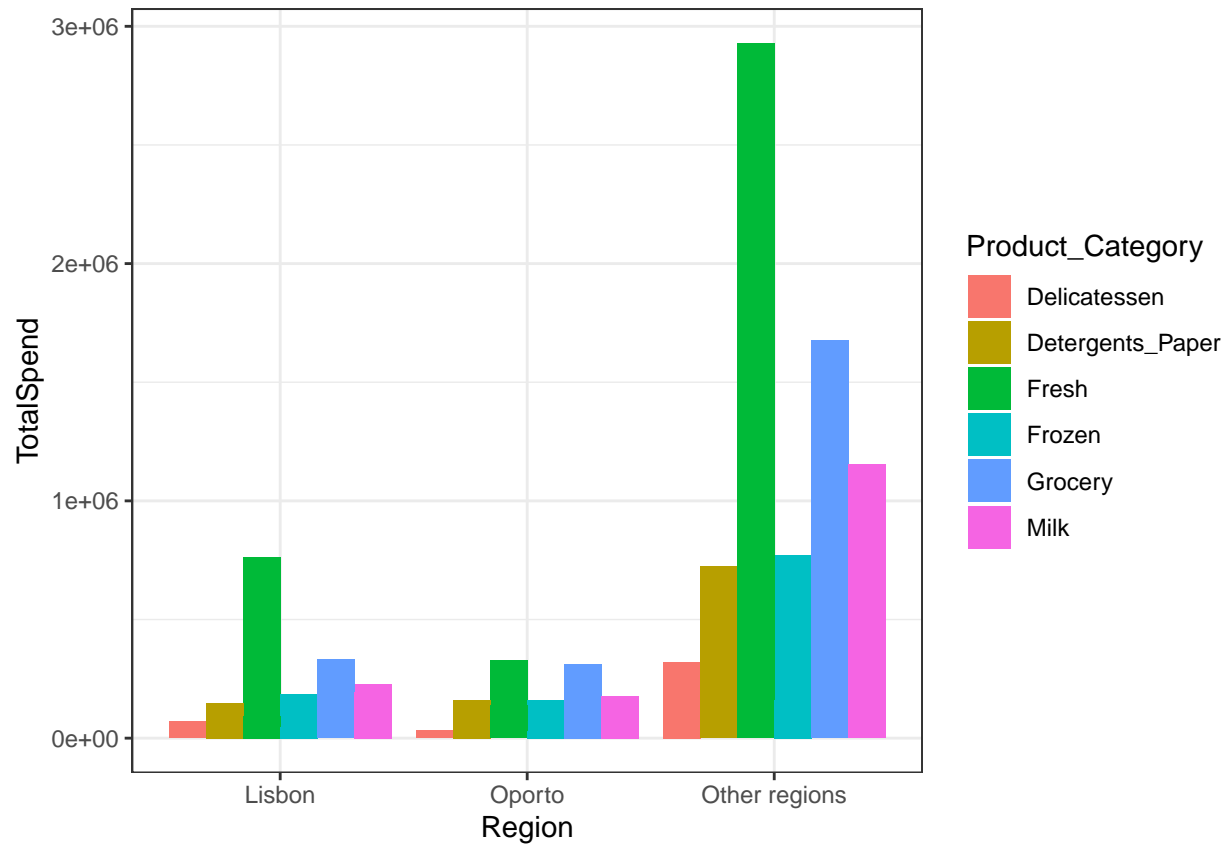
```

## TOTAL SPEND PATTERN
# Across Channels
ggplot(temp2, aes(Channel, TotalSpend, fill = Product_Category)) +
  geom_bar(stat = 'identity', position = 'dodge') +
  theme_bw()

```



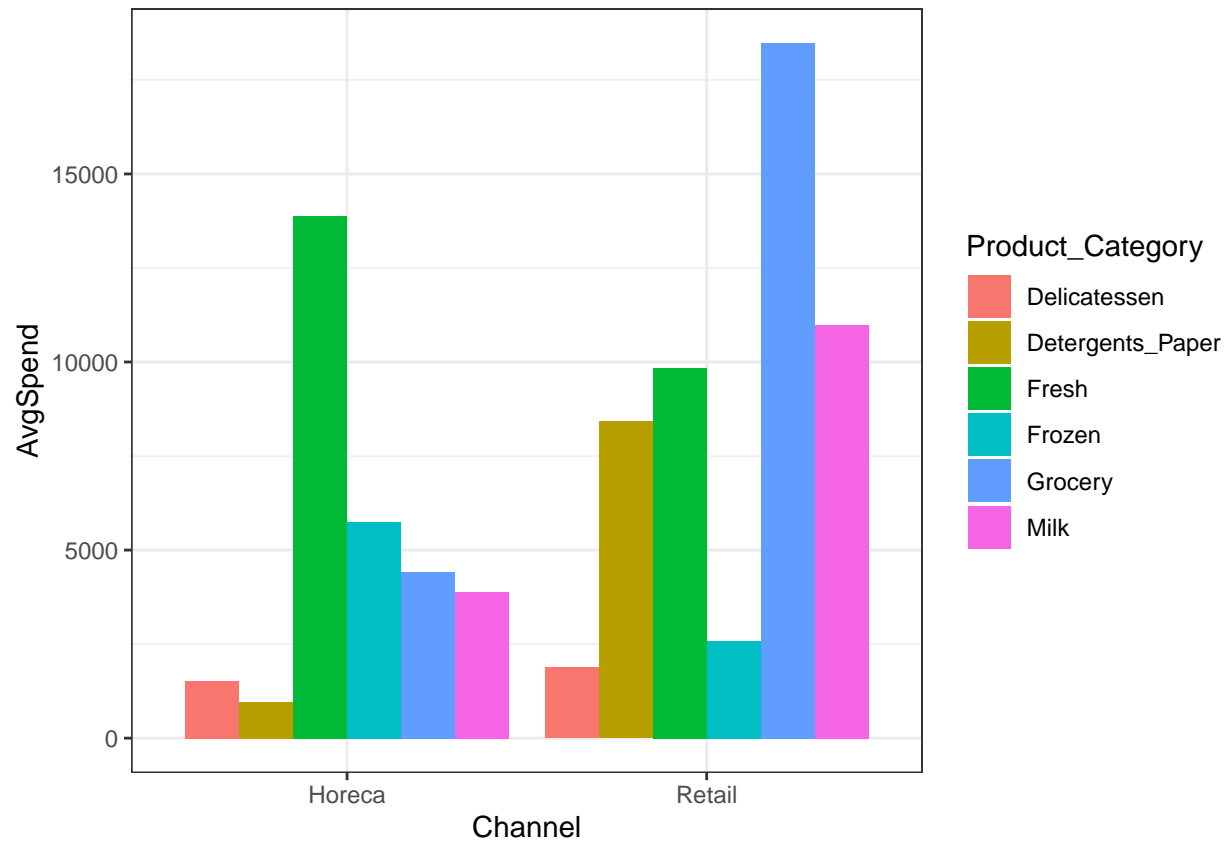
```
# Across Regions
ggplot(temp2, aes(Region, TotalSpend, fill = Product_Category)) +
  geom_bar(stat = 'identity', position = 'dodge') +
  theme_bw()
```



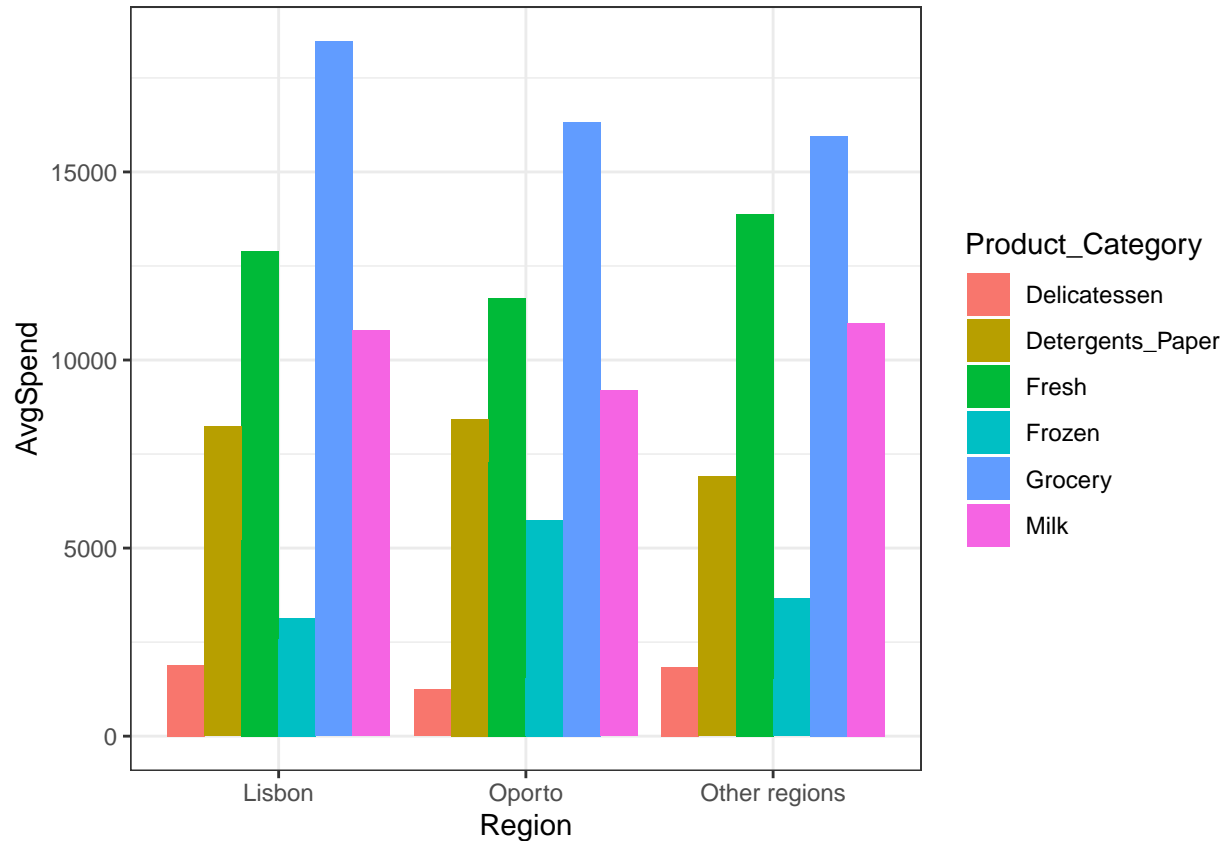
```
## AVERAGE SPEND PATTERN
```

```
# Across Channels
```

```
ggplot(temp2, aes(Channel, AvgSpend, fill = Product_Category)) +  
  geom_bar(stat = 'identity', position = 'dodge') +  
  theme_bw()
```



```
# Across Regions  
ggplot(temp2, aes(Region, AvgSpend, fill = Product_Category)) +  
  geom_bar(stat = 'identity', position = 'dodge') +  
  theme_bw()
```



Although our data come with the same scale, they still have relatively large gaps in amounts. Therefore, we applied normalization to preprocess our data and calculated the distance matrix using euclidean method accordingly.

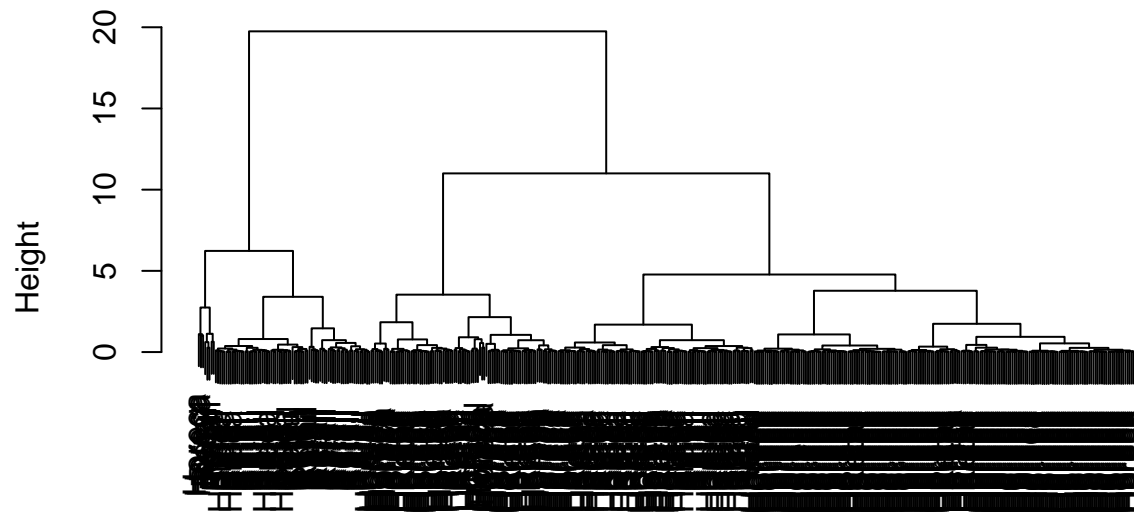
```
# Normalize the raw data

normalize = function(x){
  return ((x - min(x))/(max(x) - min(x)))
}
wholesale_norm = wholesale %>% mutate_at(c(3:8), normalize)
distance_matrix = dist(wholesale_norm[,3:8], method = "euclidean")
```

To decide on the optimized clusters, we first used hierarchical clustering analysis with Ward distance measurement method to generate a dendrogram. When deciding on the number of clusters, we first looked at the case where  $k = 5$ . As shown in the dendrogram, there would be a very small cluster generated under 5-cluster scenario, which we considered less meaningful to analyze. Therefore, we eventually landed on 3 clusters.

```
# Calculate hierarchial clustering and display dendrogram
h_clust = hclust(distance_matrix, method = "ward.D")
plot(h_clust, labels = (wholesale_norm$Channel))
```

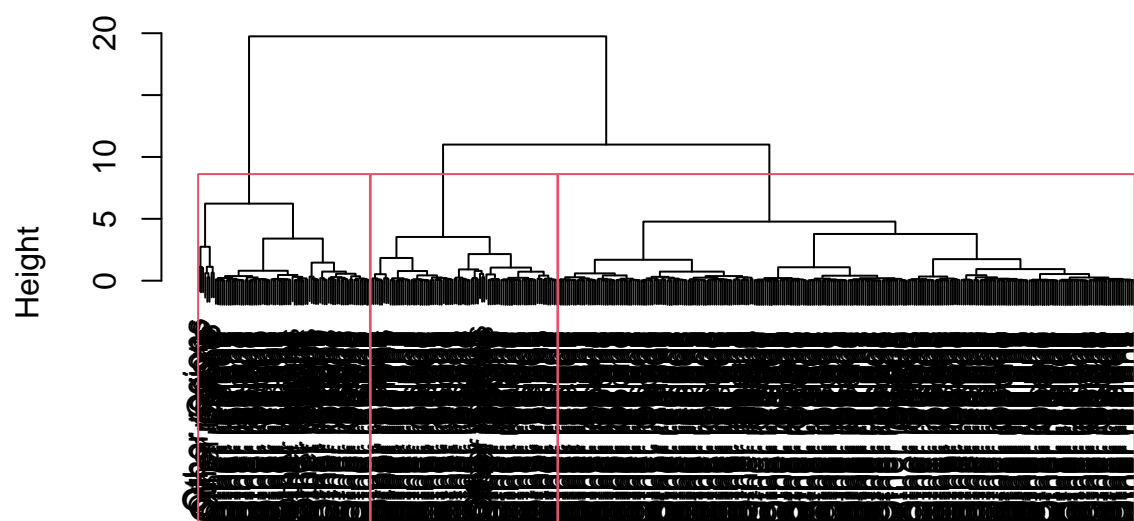
## Cluster Dendrogram



distance\_matrix  
hclust (\*, "ward.D")

```
plot(h_clust, labels = (wholesale_norm$Region))  
rect.hclust(h_clust, k = 3)
```

## Cluster Dendrogram

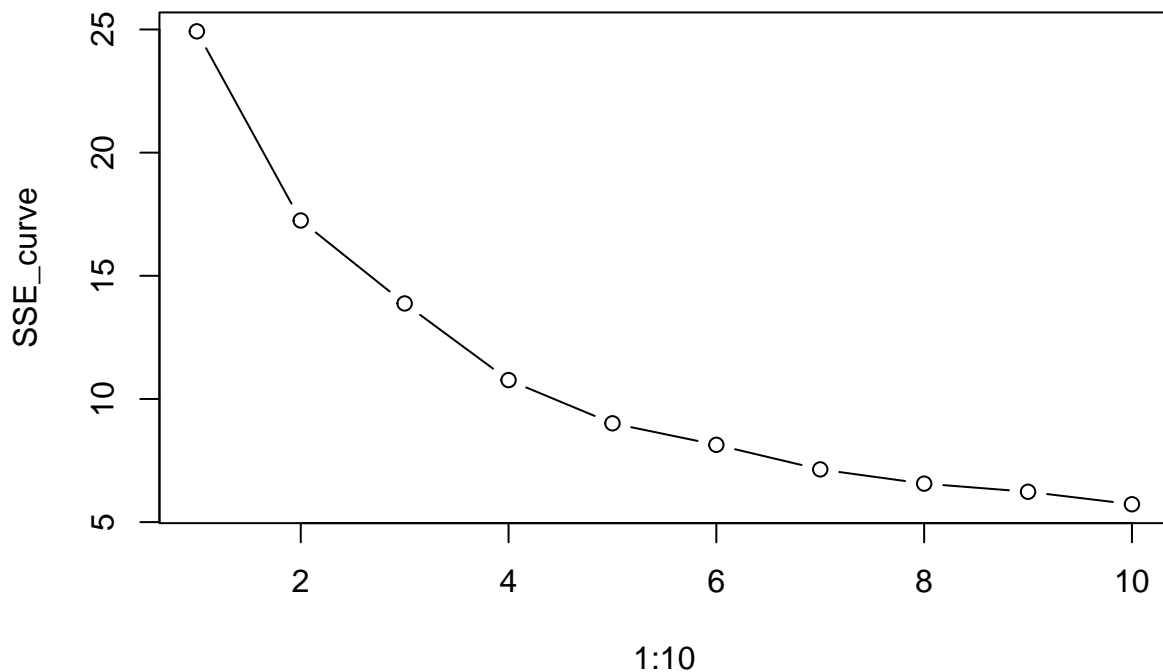


```
distance_matrix  
hclust (*, "ward.D")
```

Then, we used the k-means method to do the clustering analysis. We plotted an SSE curve to look into the proper number of clusters. Looking into the plot, we noticed a relatively larger SSE gain when  $k = 3$  and  $k = 5$ , similar to the findings in hierarchical clustering analysis.

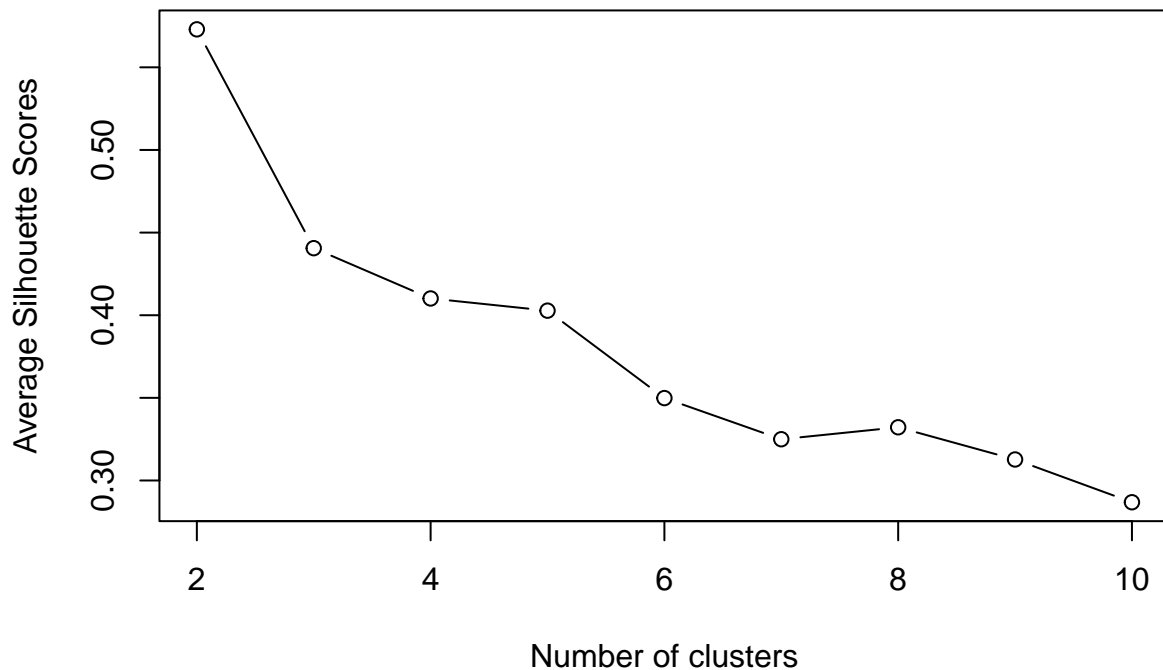
```
SSE_curve <- c()  
for (n in 1:10) {  
  kcluster = kmeans(wholesale_norm[,3:8], n)  
  sse = kcluster$tot.withinss  
  SSE_curve[n] = sse  
}  
# plot SSE against number of clusters  
plot(1:10, SSE_curve, type = "b")
```





Then, we tested through different  $k$  to calculate the mean silhouette score for each, as an additional support which shows the goodness of the clustering. It ends up that when  $k = 3$ , we can get a relatively good clustering result.

```
#test optimal k
silhouette_score <- function(k){
  kcluster <- kmeans(wholesale_norm[,3:8], centers = k, nstart = 25)
  ss <- silhouette(kcluster$cluster, dist = distance_matrix)
  mean(ss[, 3])
}
k_test = 2:10
avg_sil_score <- sapply(k_test, silhouette_score)
plot(k_test, type='b', avg_sil_score, xlab='Number of clusters', ylab='Average Silhouette Scores')
```



We then looked into the cluster distribution under the two different scenarios. Similar to hierarchical clustering analysis, we found some of the clusters under the 5-cluster scenario is too small to analyze. Therefore, we also landed on 3 clusters under the kmeans method.

#### ## K-means with 5 clusters

```
kcluster1 = kmeans(wholesale_norm[,3:8], centers = 5)
cluster1 = wholesale_norm %>% mutate(Cluster = kcluster1$cluster)
cluster1 %>% group_by(Cluster) %>% summarise(count = n())
```

```
## # A tibble: 5 x 2
##   Cluster count
##   <int> <int>
## 1     1    101
## 2     2    257
## 3     3     28
## 4     4     49
## 5     5      5
```

*# Observation - there are clusters with only 3 and 10 data points, which might not be helpful in general*

#### ## K-means with 3 clusters

```
kcluster2 = kmeans(wholesale_norm[,3:8], centers = 3)
cluster2 = wholesale_norm %>% mutate(Cluster = kcluster2$cluster)
cluster2 %>% group_by(Cluster) %>% summarise(count = n())
```

```
## # A tibble: 3 x 2
```

```
##      Cluster count
##      <int> <int>
## 1         1   338
## 2         2    61
## 3         3    41
```

The next step would be choosing between the two clustering results from the two analysis methods. With a glimpse in the clustering centroids, we found that the two clustering results showed similar centroids, which we considered as a good sign.

```
# View h-clusters in normalized data
wholesale_norm$cluster = cutree(h_clust, k=3)
wholesale_norm %>% group_by(cluster) %>% summarise_at(c(3:8), mean) %>% mutate_if(is.numeric, round, digits=2)
```

```
## # A tibble: 3 x 7
##   cluster  Fresh    Milk Grocery Frozen Detergents_Paper Delicatessen
##   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>         <dbl>         <dbl>
## 1       1 0.0727 0.0439 0.0471 0.0318         0.0309         0.0218
## 2       2 0.243 0.0740 0.0627 0.110         0.0252         0.0480
## 3       3 0.0737 0.198 0.240 0.0464         0.252         0.0473
```

```
# K-Means clustering (centroids)
kcluster2 = kmeans(wholesale_norm[,3:8], centers = 3)
kcluster2$centers
```

```
##           Fresh           Milk           Grocery           Frozen Detergents_Paper Delicatessen
## 1 0.07297318 0.05468315 0.06047771 0.03993169         0.04513443 0.02316561
## 2 0.31612924 0.08086923 0.07042464 0.11874307         0.02650803 0.06815213
## 3 0.07612741 0.26781048 0.31604316 0.03156052         0.34516819 0.04830937
```

We then calculated the Silhouette coefficients to evaluate the two clustering results. The kmeans method gives a better coefficient than the hclust. We then used the k-means clustering result for the following analysis.

```
# Silhouette coefficient - k means
library(cluster)
sc = silhouette(kcluster2$cluster, dist = distance_matrix)
summary(sc)
```

```
## Silhouette of 440 units in 3 clusters from silhouette.default(x = kcluster2$cluster, dist = distance_matrix)
## Cluster sizes and average silhouette widths:
##           338           61           41
## 0.5320006 0.0961455 0.1993003
## Individual silhouette widths:
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## -0.1810 0.2999 0.5179 0.4406 0.6217 0.6735
```

```
# Silhouette coefficient - hclust
library(cluster)
sc = silhouette(wholesale_norm$cluster, dist = distance_matrix)
summary(sc)
```

```
## Silhouette of 440 units in 3 clusters from silhouette.default(x = wholesale_norm$cluster, dist = dist)
## Cluster sizes and average silhouette widths:
##      271      88      81
## 0.53366522 0.03492611 0.03904118
## Individual silhouette widths:
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## -0.4102  0.1738  0.4277  0.3429  0.5934  0.6683
```

We matched the clusters back to the original dataset.

```
wholesale_clustered = wholesale %>% mutate(Cluster = kcluster2$cluster)
wholesale_clustered
```

	Channel	Region	Fresh	Milk	Grocery	Frozen	Detergents_Paper
## 1	Retail	Other regions	12669	9656	7561	214	2674
## 2	Retail	Other regions	7057	9810	9568	1762	3293
## 3	Retail	Other regions	6353	8808	7684	2405	3516
## 4	Horeca	Other regions	13265	1196	4221	6404	507
## 5	Retail	Other regions	22615	5410	7198	3915	1777
## 6	Retail	Other regions	9413	8259	5126	666	1795
## 7	Retail	Other regions	12126	3199	6975	480	3140
## 8	Retail	Other regions	7579	4956	9426	1669	3321
## 9	Horeca	Other regions	5963	3648	6192	425	1716
## 10	Retail	Other regions	6006	11093	18881	1159	7425
## 11	Retail	Other regions	3366	5403	12974	4400	5977
## 12	Retail	Other regions	13146	1124	4523	1420	549
## 13	Retail	Other regions	31714	12319	11757	287	3881
## 14	Retail	Other regions	21217	6208	14982	3095	6707
## 15	Retail	Other regions	24653	9465	12091	294	5058
## 16	Horeca	Other regions	10253	1114	3821	397	964
## 17	Retail	Other regions	1020	8816	12121	134	4508
## 18	Horeca	Other regions	5876	6157	2933	839	370
## 19	Retail	Other regions	18601	6327	10099	2205	2767
## 20	Horeca	Other regions	7780	2495	9464	669	2518
## 21	Retail	Other regions	17546	4519	4602	1066	2259
## 22	Horeca	Other regions	5567	871	2010	3383	375
## 23	Horeca	Other regions	31276	1917	4469	9408	2381
## 24	Retail	Other regions	26373	36423	22019	5154	4337
## 25	Retail	Other regions	22647	9776	13792	2915	4482
## 26	Retail	Other regions	16165	4230	7595	201	4003
## 27	Horeca	Other regions	9898	961	2861	3151	242
## 28	Horeca	Other regions	14276	803	3045	485	100
## 29	Retail	Other regions	4113	20484	25957	1158	8604
## 30	Horeca	Other regions	43088	2100	2609	1200	1107
## 31	Horeca	Other regions	18815	3610	11107	1148	2134
## 32	Horeca	Other regions	2612	4339	3133	2088	820
## 33	Horeca	Other regions	21632	1318	2886	266	918
## 34	Horeca	Other regions	29729	4786	7326	6130	361
## 35	Horeca	Other regions	1502	1979	2262	425	483
## 36	Retail	Other regions	688	5491	11091	833	4239
## 37	Horeca	Other regions	29955	4362	5428	1729	862
## 38	Retail	Other regions	15168	10556	12477	1920	6506
## 39	Retail	Other regions	4591	15729	16709	33	6956

## 40	Horeca Other regions	56159	555	902	10002	212
## 41	Horeca Other regions	24025	4332	4757	9510	1145
## 42	Horeca Other regions	19176	3065	5956	2033	2575
## 43	Retail Other regions	10850	7555	14961	188	6899
## 44	Retail Other regions	630	11095	23998	787	9529
## 45	Retail Other regions	9670	7027	10471	541	4618
## 46	Retail Other regions	5181	22044	21531	1740	7353
## 47	Retail Other regions	3103	14069	21955	1668	6792
## 48	Retail Other regions	44466	54259	55571	7782	24171
## 49	Retail Other regions	11519	6152	10868	584	5121
## 50	Retail Other regions	4967	21412	28921	1798	13583
## 51	Horeca Other regions	6269	1095	1980	3860	609
## 52	Horeca Other regions	3347	4051	6996	239	1538
## 53	Retail Other regions	40721	3916	5876	532	2587
## 54	Retail Other regions	491	10473	11532	744	5611
## 55	Horeca Other regions	27329	1449	1947	2436	204
## 56	Horeca Other regions	5264	3683	5005	1057	2024
## 57	Retail Other regions	4098	29892	26866	2616	17740
## 58	Retail Other regions	5417	9933	10487	38	7572
## 59	Horeca Other regions	13779	1970	1648	596	227
## 60	Horeca Other regions	6137	5360	8040	129	3084
## 61	Retail Other regions	8590	3045	7854	96	4095
## 62	Retail Other regions	35942	38369	59598	3254	26701
## 63	Retail Other regions	7823	6245	6544	4154	4074
## 64	Retail Other regions	9396	11601	15775	2896	7677
## 65	Horeca Other regions	4760	1227	3250	3724	1247
## 66	Retail Other regions	85	20959	45828	36	24231
## 67	Horeca Other regions	9	1534	7417	175	3468
## 68	Retail Other regions	19913	6759	13462	1256	5141
## 69	Horeca Other regions	2446	7260	3993	5870	788
## 70	Horeca Other regions	8352	2820	1293	779	656
## 71	Horeca Other regions	16705	2037	3202	10643	116
## 72	Horeca Other regions	18291	1266	21042	5373	4173
## 73	Horeca Other regions	4420	5139	2661	8872	1321
## 74	Retail Other regions	19899	5332	8713	8132	764
## 75	Retail Other regions	8190	6343	9794	1285	1901
## 76	Horeca Other regions	20398	1137	3	4407	3
## 77	Horeca Other regions	717	3587	6532	7530	529
## 78	Retail Other regions	12205	12697	28540	869	12034
## 79	Horeca Other regions	10766	1175	2067	2096	301
## 80	Horeca Other regions	1640	3259	3655	868	1202
## 81	Horeca Other regions	7005	829	3009	430	610
## 82	Retail Other regions	219	9540	14403	283	7818
## 83	Retail Other regions	10362	9232	11009	737	3537
## 84	Horeca Other regions	20874	1563	1783	2320	550
## 85	Retail Other regions	11867	3327	4814	1178	3837
## 86	Retail Other regions	16117	46197	92780	1026	40827
## 87	Retail Other regions	22925	73498	32114	987	20070
## 88	Horeca Other regions	43265	5025	8117	6312	1579
## 89	Horeca Other regions	7864	542	4042	9735	165
## 90	Horeca Other regions	24904	3836	5330	3443	454
## 91	Horeca Other regions	11405	596	1638	3347	69
## 92	Horeca Other regions	12754	2762	2530	8693	627
## 93	Retail Other regions	9198	27472	32034	3232	18906

## 94	Horeca Other regions	11314	3090	2062	35009	71
## 95	Retail Other regions	5626	12220	11323	206	5038
## 96	Horeca Other regions	3	2920	6252	440	223
## 97	Retail Other regions	23	2616	8118	145	3874
## 98	Horeca Other regions	403	254	610	774	54
## 99	Horeca Other regions	503	112	778	895	56
## 100	Horeca Other regions	9658	2182	1909	5639	215
## 101	Retail Other regions	11594	7779	12144	3252	8035
## 102	Retail Other regions	1420	10810	16267	1593	6766
## 103	Retail Other regions	2932	6459	7677	2561	4573
## 104	Horeca Other regions	56082	3504	8906	18028	1480
## 105	Horeca Other regions	14100	2132	3445	1336	1491
## 106	Horeca Other regions	15587	1014	3970	910	139
## 107	Retail Other regions	1454	6337	10704	133	6830
## 108	Retail Other regions	8797	10646	14886	2471	8969
## 109	Retail Other regions	1531	8397	6981	247	2505
## 110	Retail Other regions	1406	16729	28986	673	836
## 111	Horeca Other regions	11818	1648	1694	2276	169
## 112	Retail Other regions	12579	11114	17569	805	6457
## 113	Horeca Other regions	19046	2770	2469	8853	483
## 114	Horeca Other regions	14438	2295	1733	3220	585
## 115	Horeca Other regions	18044	1080	2000	2555	118
## 116	Horeca Other regions	11134	793	2988	2715	276
## 117	Horeca Other regions	11173	2521	3355	1517	310
## 118	Horeca Other regions	6990	3880	5380	1647	319
## 119	Horeca Other regions	20049	1891	2362	5343	411
## 120	Horeca Other regions	8258	2344	2147	3896	266
## 121	Horeca Other regions	17160	1200	3412	2417	174
## 122	Horeca Other regions	4020	3234	1498	2395	264
## 123	Horeca Other regions	12212	201	245	1991	25
## 124	Retail Other regions	11170	10769	8814	2194	1976
## 125	Horeca Other regions	36050	1642	2961	4787	500
## 126	Horeca Other regions	76237	3473	7102	16538	778
## 127	Horeca Other regions	19219	1840	1658	8195	349
## 128	Retail Other regions	21465	7243	10685	880	2386
## 129	Horeca Other regions	140	8847	3823	142	1062
## 130	Horeca Other regions	42312	926	1510	1718	410
## 131	Horeca Other regions	7149	2428	699	6316	395
## 132	Horeca Other regions	2101	589	314	346	70
## 133	Horeca Other regions	14903	2032	2479	576	955
## 134	Horeca Other regions	9434	1042	1235	436	256
## 135	Horeca Other regions	7388	1882	2174	720	47
## 136	Horeca Other regions	6300	1289	2591	1170	199
## 137	Horeca Other regions	4625	8579	7030	4575	2447
## 138	Horeca Other regions	3087	8080	8282	661	721
## 139	Horeca Other regions	13537	4257	5034	155	249
## 140	Horeca Other regions	5387	4979	3343	825	637
## 141	Horeca Other regions	17623	4280	7305	2279	960
## 142	Horeca Other regions	30379	13252	5189	321	51
## 143	Horeca Other regions	37036	7152	8253	2995	20
## 144	Horeca Other regions	10405	1596	1096	8425	399
## 145	Horeca Other regions	18827	3677	1988	118	516
## 146	Retail Other regions	22039	8384	34792	42	12591
## 147	Horeca Other regions	7769	1936	2177	926	73

## 148	Horeca Other regions	9203	3373	2707	1286	1082
## 149	Horeca Other regions	5924	584	542	4052	283
## 150	Horeca Other regions	31812	1433	1651	800	113
## 151	Horeca Other regions	16225	1825	1765	853	170
## 152	Horeca Other regions	1289	3328	2022	531	255
## 153	Horeca Other regions	18840	1371	3135	3001	352
## 154	Horeca Other regions	3463	9250	2368	779	302
## 155	Horeca Other regions	622	55	137	75	7
## 156	Retail Other regions	1989	10690	19460	233	11577
## 157	Retail Other regions	3830	5291	14855	317	6694
## 158	Horeca Other regions	17773	1366	2474	3378	811
## 159	Retail Other regions	2861	6570	9618	930	4004
## 160	Retail Other regions	355	7704	14682	398	8077
## 161	Retail Other regions	1725	3651	12822	824	4424
## 162	Horeca Other regions	12434	540	283	1092	3
## 163	Horeca Other regions	15177	2024	3810	2665	232
## 164	Retail Other regions	5531	15726	26870	2367	13726
## 165	Retail Other regions	5224	7603	8584	2540	3674
## 166	Retail Other regions	15615	12653	19858	4425	7108
## 167	Retail Other regions	4822	6721	9170	993	4973
## 168	Horeca Other regions	2926	3195	3268	405	1680
## 169	Horeca Other regions	5809	735	803	1393	79
## 170	Horeca Other regions	5414	717	2155	2399	69
## 171	Retail Other regions	260	8675	13430	1116	7015
## 172	Retail Other regions	200	25862	19816	651	8773
## 173	Horeca Other regions	955	5479	6536	333	2840
## 174	Retail Other regions	514	7677	19805	937	9836
## 175	Horeca Other regions	286	1208	5241	2515	153
## 176	Retail Other regions	2343	7845	11874	52	4196
## 177	Horeca Other regions	45640	6958	6536	7368	1532
## 178	Horeca Other regions	12759	7330	4533	1752	20
## 179	Horeca Other regions	11002	7075	4945	1152	120
## 180	Horeca Other regions	3157	4888	2500	4477	273
## 181	Horeca Other regions	12356	6036	8887	402	1382
## 182	Horeca Other regions	112151	29627	18148	16745	4948
## 183	Horeca Other regions	694	8533	10518	443	6907
## 184	Horeca Other regions	36847	43950	20170	36534	239
## 185	Horeca Other regions	327	918	4710	74	334
## 186	Horeca Other regions	8170	6448	1139	2181	58
## 187	Horeca Other regions	3009	521	854	3470	949
## 188	Horeca Other regions	2438	8002	9819	6269	3459
## 189	Retail Other regions	8040	7639	11687	2758	6839
## 190	Retail Other regions	834	11577	11522	275	4027
## 191	Horeca Other regions	16936	6250	1981	7332	118
## 192	Horeca Other regions	13624	295	1381	890	43
## 193	Horeca Other regions	5509	1461	2251	547	187
## 194	Retail Other regions	180	3485	20292	959	5618
## 195	Horeca Other regions	7107	1012	2974	806	355
## 196	Horeca Other regions	17023	5139	5230	7888	330
## 197	Horeca Lisbon	30624	7209	4897	18711	763
## 198	Retail Lisbon	2427	7097	10391	1127	4314
## 199	Horeca Lisbon	11686	2154	6824	3527	592
## 200	Horeca Lisbon	9670	2280	2112	520	402
## 201	Retail Lisbon	3067	13240	23127	3941	9959

## 202	Retail	Lisbon	4484	14399	24708	3549	14235
## 203	Horeca	Lisbon	25203	11487	9490	5065	284
## 204	Horeca	Lisbon	583	685	2216	469	954
## 205	Horeca	Lisbon	1956	891	5226	1383	5
## 206	Retail	Lisbon	1107	11711	23596	955	9265
## 207	Horeca	Lisbon	6373	780	950	878	288
## 208	Retail	Lisbon	2541	4737	6089	2946	5316
## 209	Horeca	Lisbon	1537	3748	5838	1859	3381
## 210	Retail	Lisbon	5550	12729	16767	864	12420
## 211	Horeca	Lisbon	18567	1895	1393	1801	244
## 212	Retail	Lisbon	12119	28326	39694	4736	19410
## 213	Horeca	Lisbon	7291	1012	2062	1291	240
## 214	Horeca	Lisbon	3317	6602	6861	1329	3961
## 215	Retail	Lisbon	2362	6551	11364	913	5957
## 216	Horeca	Lisbon	2806	10765	15538	1374	5828
## 217	Retail	Lisbon	2532	16599	36486	179	13308
## 218	Horeca	Lisbon	18044	1475	2046	2532	130
## 219	Retail	Lisbon	18	7504	15205	1285	4797
## 220	Horeca	Lisbon	4155	367	1390	2306	86
## 221	Horeca	Lisbon	14755	899	1382	1765	56
## 222	Horeca	Lisbon	5396	7503	10646	91	4167
## 223	Horeca	Lisbon	5041	1115	2856	7496	256
## 224	Retail	Lisbon	2790	2527	5265	5612	788
## 225	Horeca	Lisbon	7274	659	1499	784	70
## 226	Horeca	Lisbon	12680	3243	4157	660	761
## 227	Retail	Lisbon	20782	5921	9212	1759	2568
## 228	Horeca	Lisbon	4042	2204	1563	2286	263
## 229	Horeca	Lisbon	1869	577	572	950	4762
## 230	Horeca	Lisbon	8656	2746	2501	6845	694
## 231	Retail	Lisbon	11072	5989	5615	8321	955
## 232	Horeca	Lisbon	2344	10678	3828	1439	1566
## 233	Horeca	Lisbon	25962	1780	3838	638	284
## 234	Horeca	Lisbon	964	4984	3316	937	409
## 235	Horeca	Lisbon	15603	2703	3833	4260	325
## 236	Horeca	Lisbon	1838	6380	2824	1218	1216
## 237	Horeca	Lisbon	8635	820	3047	2312	415
## 238	Horeca	Lisbon	18692	3838	593	4634	28
## 239	Horeca	Lisbon	7363	475	585	1112	72
## 240	Horeca	Lisbon	47493	2567	3779	5243	828
## 241	Horeca	Lisbon	22096	3575	7041	11422	343
## 242	Horeca	Lisbon	24929	1801	2475	2216	412
## 243	Horeca	Lisbon	18226	659	2914	3752	586
## 244	Horeca	Lisbon	11210	3576	5119	561	1682
## 245	Horeca	Lisbon	6202	7775	10817	1183	3143
## 246	Retail	Lisbon	3062	6154	13916	230	8933
## 247	Horeca	Lisbon	8885	2428	1777	1777	430
## 248	Horeca	Lisbon	13569	346	489	2077	44
## 249	Horeca	Lisbon	15671	5279	2406	559	562
## 250	Horeca	Lisbon	8040	3795	2070	6340	918
## 251	Horeca	Lisbon	3191	1993	1799	1730	234
## 252	Retail	Lisbon	6134	23133	33586	6746	18594
## 253	Horeca	Lisbon	6623	1860	4740	7683	205
## 254	Horeca	Lisbon	29526	7961	16966	432	363
## 255	Horeca	Lisbon	10379	17972	4748	4686	1547



##	256	Horeca	Lisbon	31614	489	1495	3242	111
##	257	Horeca	Lisbon	11092	5008	5249	453	392
##	258	Horeca	Lisbon	8475	1931	1883	5004	3593
##	259	Horeca	Lisbon	56083	4563	2124	6422	730
##	260	Horeca	Lisbon	53205	4959	7336	3012	967
##	261	Horeca	Lisbon	9193	4885	2157	327	780
##	262	Horeca	Lisbon	7858	1110	1094	6818	49
##	263	Horeca	Lisbon	23257	1372	1677	982	429
##	264	Horeca	Lisbon	2153	1115	6684	4324	2894
##	265	Retail	Lisbon	1073	9679	15445	61	5980
##	266	Horeca	Lisbon	5909	23527	13699	10155	830
##	267	Retail	Lisbon	572	9763	22182	2221	4882
##	268	Horeca	Lisbon	20893	1222	2576	3975	737
##	269	Retail	Lisbon	11908	8053	19847	1069	6374
##	270	Horeca	Lisbon	15218	258	1138	2516	333
##	271	Horeca	Lisbon	4720	1032	975	5500	197
##	272	Horeca	Lisbon	2083	5007	1563	1120	147
##	273	Horeca	Lisbon	514	8323	6869	529	93
##	274	Horeca	Other regions	36817	3045	1493	4802	210
##	275	Horeca	Other regions	894	1703	1841	744	759
##	276	Horeca	Other regions	680	1610	223	862	96
##	277	Horeca	Other regions	27901	3749	6964	4479	603
##	278	Horeca	Other regions	9061	829	683	16919	621
##	279	Horeca	Other regions	11693	2317	2543	5845	274
##	280	Retail	Other regions	17360	6200	9694	1293	3620
##	281	Horeca	Other regions	3366	2884	2431	977	167
##	282	Retail	Other regions	12238	7108	6235	1093	2328
##	283	Horeca	Other regions	49063	3965	4252	5970	1041
##	284	Horeca	Other regions	25767	3613	2013	10303	314
##	285	Horeca	Other regions	68951	4411	12609	8692	751
##	286	Horeca	Other regions	40254	640	3600	1042	436
##	287	Horeca	Other regions	7149	2247	1242	1619	1226
##	288	Horeca	Other regions	15354	2102	2828	8366	386
##	289	Horeca	Other regions	16260	594	1296	848	445
##	290	Horeca	Other regions	42786	286	471	1388	32
##	291	Horeca	Other regions	2708	2160	2642	502	965
##	292	Horeca	Other regions	6022	3354	3261	2507	212
##	293	Horeca	Other regions	2838	3086	4329	3838	825
##	294	Retail	Oporto	3996	11103	12469	902	5952
##	295	Horeca	Oporto	21273	2013	6550	909	811
##	296	Retail	Oporto	7588	1897	5234	417	2208
##	297	Horeca	Oporto	19087	1304	3643	3045	710
##	298	Retail	Oporto	8090	3199	6986	1455	3712
##	299	Retail	Oporto	6758	4560	9965	934	4538
##	300	Horeca	Oporto	444	879	2060	264	290
##	301	Retail	Oporto	16448	6243	6360	824	2662
##	302	Retail	Oporto	5283	13316	20399	1809	8752
##	303	Retail	Oporto	2886	5302	9785	364	6236
##	304	Retail	Oporto	2599	3688	13829	492	10069
##	305	Retail	Oporto	161	7460	24773	617	11783
##	306	Retail	Oporto	243	12939	8852	799	3909
##	307	Retail	Oporto	6468	12867	21570	1840	7558
##	308	Horeca	Oporto	17327	2374	2842	1149	351
##	309	Horeca	Oporto	6987	1020	3007	416	257

## 310	Retail	Oporto	918	20655	13567	1465	6846
## 311	Horeca	Oporto	7034	1492	2405	12569	299
## 312	Horeca	Oporto	29635	2335	8280	3046	371
## 313	Retail	Oporto	2137	3737	19172	1274	17120
## 314	Horeca	Oporto	9784	925	2405	4447	183
## 315	Horeca	Oporto	10617	1795	7647	1483	857
## 316	Retail	Oporto	1479	14982	11924	662	3891
## 317	Horeca	Oporto	7127	1375	2201	2679	83
## 318	Horeca	Oporto	1182	3088	6114	978	821
## 319	Horeca	Oporto	11800	2713	3558	2121	706
## 320	Retail	Oporto	9759	25071	17645	1128	12408
## 321	Horeca	Oporto	1774	3696	2280	514	275
## 322	Horeca	Oporto	9155	1897	5167	2714	228
## 323	Horeca	Oporto	15881	713	3315	3703	1470
## 324	Horeca	Oporto	13360	944	11593	915	1679
## 325	Horeca	Oporto	25977	3587	2464	2369	140
## 326	Horeca	Oporto	32717	16784	13626	60869	1272
## 327	Horeca	Oporto	4414	1610	1431	3498	387
## 328	Horeca	Oporto	542	899	1664	414	88
## 329	Horeca	Oporto	16933	2209	3389	7849	210
## 330	Horeca	Oporto	5113	1486	4583	5127	492
## 331	Horeca	Oporto	9790	1786	5109	3570	182
## 332	Retail	Oporto	11223	14881	26839	1234	9606
## 333	Horeca	Oporto	22321	3216	1447	2208	178
## 334	Retail	Oporto	8565	4980	67298	131	38102
## 335	Retail	Oporto	16823	928	2743	11559	332
## 336	Retail	Oporto	27082	6817	10790	1365	4111
## 337	Horeca	Oporto	13970	1511	1330	650	146
## 338	Horeca	Oporto	9351	1347	2611	8170	442
## 339	Horeca	Oporto	3	333	7021	15601	15
## 340	Horeca	Oporto	2617	1188	5332	9584	573
## 341	Retail Other regions		381	4025	9670	388	7271
## 342	Retail Other regions		2320	5763	11238	767	5162
## 343	Horeca Other regions		255	5758	5923	349	4595
## 344	Retail Other regions		1689	6964	26316	1456	15469
## 345	Horeca Other regions		3043	1172	1763	2234	217
## 346	Horeca Other regions		1198	2602	8335	402	3843
## 347	Retail Other regions		2771	6939	15541	2693	6600
## 348	Retail Other regions		27380	7184	12311	2809	4621
## 349	Horeca Other regions		3428	2380	2028	1341	1184
## 350	Retail Other regions		5981	14641	20521	2005	12218
## 351	Horeca Other regions		3521	1099	1997	1796	173
## 352	Retail Other regions		1210	10044	22294	1741	12638
## 353	Horeca Other regions		608	1106	1533	830	90
## 354	Retail Other regions		117	6264	21203	228	8682
## 355	Horeca Other regions		14039	7393	2548	6386	1333
## 356	Horeca Other regions		190	727	2012	245	184
## 357	Horeca Other regions		22686	134	218	3157	9
## 358	Retail Other regions		37	1275	22272	137	6747
## 359	Horeca Other regions		759	18664	1660	6114	536
## 360	Horeca Other regions		796	5878	2109	340	232
## 361	Horeca Other regions		19746	2872	2006	2601	468
## 362	Horeca Other regions		4734	607	864	1206	159
## 363	Horeca Other regions		2121	1601	2453	560	179

## 364	Horeca Other regions	4627	997	4438	191	1335
## 365	Horeca Other regions	2615	873	1524	1103	514
## 366	Retail Other regions	4692	6128	8025	1619	4515
## 367	Horeca Other regions	9561	2217	1664	1173	222
## 368	Horeca Other regions	3477	894	534	1457	252
## 369	Horeca Other regions	22335	1196	2406	2046	101
## 370	Horeca Other regions	6211	337	683	1089	41
## 371	Retail Other regions	39679	3944	4955	1364	523
## 372	Horeca Other regions	20105	1887	1939	8164	716
## 373	Horeca Other regions	3884	3801	1641	876	397
## 374	Retail Other regions	15076	6257	7398	1504	1916
## 375	Horeca Other regions	6338	2256	1668	1492	311
## 376	Horeca Other regions	5841	1450	1162	597	476
## 377	Retail Other regions	3136	8630	13586	5641	4666
## 378	Horeca Other regions	38793	3154	2648	1034	96
## 379	Horeca Other regions	3225	3294	1902	282	68
## 380	Retail Other regions	4048	5164	10391	130	813
## 381	Horeca Other regions	28257	944	2146	3881	600
## 382	Horeca Other regions	17770	4591	1617	9927	246
## 383	Horeca Other regions	34454	7435	8469	2540	1711
## 384	Horeca Other regions	1821	1364	3450	4006	397
## 385	Horeca Other regions	10683	21858	15400	3635	282
## 386	Horeca Other regions	11635	922	1614	2583	192
## 387	Horeca Other regions	1206	3620	2857	1945	353
## 388	Horeca Other regions	20918	1916	1573	1960	231
## 389	Horeca Other regions	9785	848	1172	1677	200
## 390	Horeca Other regions	9385	1530	1422	3019	227
## 391	Horeca Other regions	3352	1181	1328	5502	311
## 392	Horeca Other regions	2647	2761	2313	907	95
## 393	Horeca Other regions	518	4180	3600	659	122
## 394	Horeca Other regions	23632	6730	3842	8620	385
## 395	Horeca Other regions	12377	865	3204	1398	149
## 396	Horeca Other regions	9602	1316	1263	2921	841
## 397	Retail Other regions	4515	11991	9345	2644	3378
## 398	Horeca Other regions	11535	1666	1428	6838	64
## 399	Horeca Other regions	11442	1032	582	5390	74
## 400	Horeca Other regions	9612	577	935	1601	469
## 401	Horeca Other regions	4446	906	1238	3576	153
## 402	Horeca Other regions	27167	2801	2128	13223	92
## 403	Horeca Other regions	26539	4753	5091	220	10
## 404	Horeca Other regions	25606	11006	4604	127	632
## 405	Horeca Other regions	18073	4613	3444	4324	914
## 406	Horeca Other regions	6884	1046	1167	2069	593
## 407	Horeca Other regions	25066	5010	5026	9806	1092
## 408	Retail Other regions	7362	12844	18683	2854	7883
## 409	Retail Other regions	8257	3880	6407	1646	2730
## 410	Horeca Other regions	8708	3634	6100	2349	2123
## 411	Horeca Other regions	6633	2096	4563	1389	1860
## 412	Horeca Other regions	2126	3289	3281	1535	235
## 413	Horeca Other regions	97	3605	12400	98	2970
## 414	Horeca Other regions	4983	4859	6633	17866	912
## 415	Horeca Other regions	5969	1990	3417	5679	1135
## 416	Retail Other regions	7842	6046	8552	1691	3540
## 417	Retail Other regions	4389	10940	10908	848	6728

## 418	Horeca Other regions	5065	5499	11055	364	3485
## 419	Retail Other regions	660	8494	18622	133	6740
## 420	Horeca Other regions	8861	3783	2223	633	1580
## 421	Horeca Other regions	4456	5266	13227	25	6818
## 422	Retail Other regions	17063	4847	9053	1031	3415
## 423	Horeca Other regions	26400	1377	4172	830	948
## 424	Retail Other regions	17565	3686	4657	1059	1803
## 425	Retail Other regions	16980	2884	12232	874	3213
## 426	Horeca Other regions	11243	2408	2593	15348	108
## 427	Horeca Other regions	13134	9347	14316	3141	5079
## 428	Horeca Other regions	31012	16687	5429	15082	439
## 429	Horeca Other regions	3047	5970	4910	2198	850
## 430	Horeca Other regions	8607	1750	3580	47	84
## 431	Horeca Other regions	3097	4230	16483	575	241
## 432	Horeca Other regions	8533	5506	5160	13486	1377
## 433	Horeca Other regions	21117	1162	4754	269	1328
## 434	Horeca Other regions	1982	3218	1493	1541	356
## 435	Horeca Other regions	16731	3922	7994	688	2371
## 436	Horeca Other regions	29703	12051	16027	13135	182
## 437	Horeca Other regions	39228	1431	764	4510	93
## 438	Retail Other regions	14531	15488	30243	437	14841
## 439	Horeca Other regions	10290	1981	2232	1038	168
## 440	Horeca Other regions	2787	1698	2510	65	477
##	Delicatessen Cluster					
## 1	1338	1				
## 2	1776	1				
## 3	7844	1				
## 4	1788	1				
## 5	5185	2				
## 6	1451	1				
## 7	545	1				
## 8	2566	1				
## 9	750	1				
## 10	2098	1				
## 11	1744	1				
## 12	497	1				
## 13	2931	2				
## 14	602	1				
## 15	2168	2				
## 16	412	1				
## 17	1080	1				
## 18	4478	1				
## 19	3181	1				
## 20	501	1				
## 21	2124	1				
## 22	569	1				
## 23	4334	2				
## 24	16523	3				
## 25	5778	2				
## 26	57	1				
## 27	833	1				
## 28	518	1				
## 29	5206	3				
## 30	823	2				

## 31	2963	1
## 32	985	1
## 33	405	1
## 34	1083	2
## 35	395	1
## 36	436	1
## 37	4626	2
## 38	714	1
## 39	433	3
## 40	2916	2
## 41	5864	2
## 42	2802	1
## 43	46	1
## 44	72	3
## 45	65	1
## 46	4985	3
## 47	1452	3
## 48	6465	3
## 49	1476	1
## 50	1163	3
## 51	2162	1
## 52	301	1
## 53	1278	2
## 54	224	1
## 55	1333	2
## 56	1130	1
## 57	1340	3
## 58	1282	1
## 59	436	1
## 60	1603	1
## 61	225	1
## 62	2017	3
## 63	964	1
## 64	1295	1
## 65	1145	1
## 66	1423	3
## 67	27	1
## 68	834	1
## 69	3095	1
## 70	144	1
## 71	1365	1
## 72	14472	2
## 73	181	1
## 74	648	1
## 75	1780	1
## 76	975	1
## 77	894	1
## 78	1009	3
## 79	167	1
## 80	1653	1
## 81	529	1
## 82	156	1
## 83	2342	1
## 84	772	1

## 85	120	1
## 86	2944	3
## 87	903	3
## 88	14351	2
## 89	46	1
## 90	3178	2
## 91	360	1
## 92	1117	1
## 93	5130	3
## 94	2698	2
## 95	244	1
## 96	709	1
## 97	217	1
## 98	63	1
## 99	132	1
## 100	323	1
## 101	3029	1
## 102	1838	1
## 103	1386	1
## 104	2498	2
## 105	548	1
## 106	1378	1
## 107	1831	1
## 108	1438	1
## 109	1236	1
## 110	3	1
## 111	1647	1
## 112	1519	1
## 113	2708	1
## 114	1561	1
## 115	1266	1
## 116	610	1
## 117	222	1
## 118	1160	1
## 119	933	1
## 120	635	1
## 121	1136	1
## 122	255	1
## 123	860	1
## 124	143	1
## 125	1621	2
## 126	918	2
## 127	483	1
## 128	2749	1
## 129	3	1
## 130	1819	2
## 131	911	1
## 132	310	1
## 133	328	1
## 134	396	1
## 135	537	1
## 136	326	1
## 137	1542	1
## 138	36	1

## 139	3271	1
## 140	929	1
## 141	2616	1
## 142	1450	2
## 143	3	2
## 144	318	1
## 145	201	1
## 146	4430	3
## 147	520	1
## 148	526	1
## 149	434	1
## 150	1440	2
## 151	1067	1
## 152	1774	1
## 153	184	1
## 154	1627	1
## 155	8	1
## 156	2153	3
## 157	3182	1
## 158	418	1
## 159	1682	1
## 160	303	1
## 161	2157	1
## 162	2233	1
## 163	610	1
## 164	446	3
## 165	238	1
## 166	2379	3
## 167	3637	1
## 168	693	1
## 169	429	1
## 170	750	1
## 171	323	1
## 172	6250	3
## 173	707	1
## 174	716	3
## 175	1442	1
## 176	1697	1
## 177	230	2
## 178	2631	1
## 179	395	1
## 180	2165	1
## 181	2794	1
## 182	8550	2
## 183	156	1
## 184	47943	2
## 185	11	1
## 186	247	1
## 187	727	1
## 188	3	1
## 189	404	1
## 190	1856	1
## 191	64	1
## 192	84	1

## 193	409	1
## 194	666	1
## 195	1142	1
## 196	1755	1
## 197	2876	2
## 198	1468	1
## 199	697	1
## 200	347	1
## 201	731	3
## 202	1681	3
## 203	6854	2
## 204	18	1
## 205	1328	1
## 206	710	3
## 207	285	1
## 208	120	1
## 209	806	1
## 210	797	3
## 211	2100	1
## 212	2870	3
## 213	1775	1
## 214	1215	1
## 215	791	1
## 216	2388	1
## 217	674	3
## 218	1158	1
## 219	6372	1
## 220	130	1
## 221	749	1
## 222	239	1
## 223	375	1
## 224	1360	1
## 225	659	1
## 226	786	1
## 227	1553	1
## 228	689	1
## 229	203	1
## 230	980	1
## 231	2137	1
## 232	490	1
## 233	834	2
## 234	7	1
## 235	2563	1
## 236	295	1
## 237	225	1
## 238	1215	1
## 239	216	1
## 240	2253	2
## 241	2564	2
## 242	1047	2
## 243	578	1
## 244	2398	1
## 245	1970	1
## 246	2784	1



## 247	610	1
## 248	659	1
## 249	572	1
## 250	291	1
## 251	710	1
## 252	5121	3
## 253	1693	1
## 254	1391	2
## 255	3265	1
## 256	615	2
## 257	373	1
## 258	987	1
## 259	3321	2
## 260	818	2
## 261	548	1
## 262	287	1
## 263	655	1
## 264	411	1
## 265	1265	1
## 266	3636	1
## 267	2563	1
## 268	3628	1
## 269	698	1
## 270	204	1
## 271	56	1
## 272	1550	1
## 273	1040	1
## 274	1824	2
## 275	1153	1
## 276	379	1
## 277	2503	2
## 278	139	1
## 279	1409	1
## 280	1721	1
## 281	1104	1
## 282	2079	1
## 283	1404	2
## 284	1384	2
## 285	2406	2
## 286	18	2
## 287	128	1
## 288	1027	1
## 289	258	1
## 290	22	2
## 291	1522	1
## 292	686	1
## 293	1060	1
## 294	741	1
## 295	1854	1
## 296	254	1
## 297	898	1
## 298	531	1
## 299	1037	1
## 300	259	1

## 301	2005	1
## 302	172	3
## 303	555	1
## 304	59	1
## 305	2410	3
## 306	211	1
## 307	1543	3
## 308	925	1
## 309	656	1
## 310	806	3
## 311	1117	1
## 312	117	2
## 313	142	3
## 314	297	1
## 315	1233	1
## 316	3508	1
## 317	1059	1
## 318	1637	1
## 319	51	1
## 320	1625	3
## 321	834	1
## 322	1113	1
## 323	229	1
## 324	573	1
## 325	1092	2
## 326	5609	2
## 327	834	1
## 328	522	1
## 329	1534	1
## 330	739	1
## 331	1043	1
## 332	1102	3
## 333	2602	1
## 334	1215	3
## 335	3486	1
## 336	2139	2
## 337	778	1
## 338	868	1
## 339	550	1
## 340	1942	1
## 341	1371	1
## 342	2158	1
## 343	1328	1
## 344	37	3
## 345	379	1
## 346	303	1
## 347	1115	1
## 348	1022	2
## 349	665	1
## 350	445	3
## 351	995	1
## 352	3137	3
## 353	195	1
## 354	1111	1

## 355	2341	1
## 356	127	1
## 357	548	1
## 358	110	1
## 359	4100	1
## 360	776	1
## 361	503	1
## 362	405	1
## 363	712	1
## 364	314	1
## 365	468	1
## 366	3105	1
## 367	447	1
## 368	342	1
## 369	558	1
## 370	296	1
## 371	2235	2
## 372	790	1
## 373	4829	1
## 374	3113	1
## 375	686	1
## 376	70	1
## 377	1426	1
## 378	1242	2
## 379	1114	1
## 380	179	1
## 381	270	2
## 382	532	1
## 383	2893	2
## 384	361	1
## 385	5120	1
## 386	1068	1
## 387	967	1
## 388	961	1
## 389	406	1
## 390	684	1
## 391	1000	1
## 392	1827	1
## 393	654	1
## 394	819	2
## 395	452	1
## 396	290	1
## 397	2213	1
## 398	743	1
## 399	247	1
## 400	375	1
## 401	1014	1
## 402	1902	2
## 403	340	2
## 404	288	2
## 405	715	1
## 406	378	1
## 407	960	2
## 408	553	3

```
## 409      344      1
## 410     5137      1
## 411     1892      1
## 412     4365      1
## 413        62      1
## 414     2435      1
## 415       290      1
## 416     1874      1
## 417       993      1
## 418     1063      1
## 419       776      1
## 420     1521      1
## 421     1393      1
## 422     1784      1
## 423     1218      2
## 424       668      1
## 425       249      1
## 426     1886      1
## 427     1894      1
## 428     1163      2
## 429       317      1
## 430     2501      1
## 431     2080      1
## 432     1498      1
## 433       395      1
## 434     1449      1
## 435       838      1
## 436     2204      2
## 437     2346      2
## 438     1867      3
## 439     2125      1
## 440        52      1
```

Looking into the spending patterns within each cluster, we found some patterns regarding both average gross spending and average categorical spending. Cluster 1: Low spending clients Cluster 2: High spending clients who prefer Milk, Grocery and Detergents Paper Cluster 3: Middle spending customer who prefer Fresh and Frozen

```
# Pivot data to create a single continuous/numerical factor using gather function
wholesale_pivot <- gather(data = wholesale_clustered, key = 'Product_Category', value = 'Spend', Fresh:)

cluster_avgspending_overall <- wholesale_pivot %>% group_by(Cluster) %>%
  summarise(TotalSpend = sum(Spend), AvgSpend = mean(Spend))
cluster_avgspending_overall
```

```
## # A tibble: 3 x 3
##   Cluster TotalSpend AvgSpend
##   <int>      <int>    <dbl>
## 1     1    7870520   3881.
## 2     2    3635143   9932.
## 3     3    3113837  12658.
```

```
cluster_avgspending_category <- wholesale_pivot %>% group_by(Cluster,Product_Category) %>%
  summarise(TotalSpend = sum(Spend),AvgSpend = mean(Spend))
```

## 'summarise()' has grouped output by 'Cluster'. You can override using the  
## '.groups' argument.

```
cluster_avgspending_category
```

```
## # A tibble: 18 x 4
## # Groups:   Cluster [3]
##   Cluster Product_Category TotalSpend AvgSpend
##   <int> <chr>           <int>    <dbl>
## 1     1 Delicatessen      376383  1114.
## 2     1 Detergents_Paper  623802  1846.
## 3     1 Fresh            2767137  8187.
## 4     1 Frozen           829656  2455.
## 5     1 Grocery          1897512  5614.
## 6     1 Milk             1376030  4071.
## 7     2 Delicatessen      199483  3270.
## 8     2 Detergents_Paper   66195  1085.
## 9     2 Fresh            2162832 35456.
## 10    2 Frozen            442238  7250.
## 11    2 Grocery           398744  6537.
## 12    2 Milk             365651  5994.
## 13    3 Delicatessen       95077  2319.
## 14    3 Detergents_Paper  577860 14094.
## 15    3 Fresh            350162  8541.
## 16    3 Frozen            79756  1945.
## 17    3 Grocery          1202306 29325.
## 18    3 Milk             808676 19724.
```

```
# Grouping data by channel and region (normalized BEFORE grouping)
```

```
# All groupings normalized
```

```
wholesale_groups_norm = wholesale_norm %>% group_by(Channel,Region) %>% summarise(Fresh_mean_norm = mean(Fresh),
wholesale_groups_norm %>% mutate_if(is.numeric, round, digits=2)
```

```
## # A tibble: 6 x 8
```

```
##   Channel Region      Fresh_mean_norm Milk_mean_norm Grocery_mean_norm
##   <chr>   <chr>           <dbl>         <dbl>         <dbl>
## 1 Horeca  Lisbon           0.12           0.05           0.04
## 2 Horeca  Oporto           0.1            0.03           0.05
## 3 Horeca  Other regions    0.12           0.05           0.04
## 4 Retail  Lisbon           0.05           0.15           0.2
## 5 Retail  Oporto           0.06           0.12           0.18
## 6 Retail  Other regions    0.09           0.15           0.17
## # i 3 more variables: Frozen_mean_norm <dbl>, Detergents_Paper_mean_norm <dbl>,
## #   Delicatessen_mean_norm <dbl>
```

```
# Group by Channel normalized
```

```
wholesale_Channel_norm = wholesale_groups_norm %>% group_by(Channel) %>% summarise(Fresh_mean_norm = mean(Fresh),
wholesale_Channel_norm %>% mutate_if(is.numeric, round, digits=2)
```

```
## # A tibble: 2 x 7
##   Channel Fresh_mean_norm Milk_mean_norm Grocery_mean_norm Frozen_mean_norm
##   <chr>         <dbl>         <dbl>         <dbl>         <dbl>
## 1 Horeca         0.11         0.04         0.04         0.07
## 2 Retail         0.07         0.14         0.18         0.03
## # i 2 more variables: Detergents_Paper_mean_norm <dbl>,
## #   Delicatessen_mean_norm <dbl>
```

```
# Group by Region normalized
```

```
wholesale_region_norm = wholesale_groups_norm %>% group_by(Region) %>% summarise(Fresh_mean_norm = mean(Fresh_mean_norm),
wholesale_region_norm %>% mutate_if(is.numeric, round, digits=2)
```

```
## # A tibble: 3 x 7
##   Region      Fresh_mean_norm Milk_mean_norm Grocery_mean_norm Frozen_mean_norm
##   <chr>         <dbl>         <dbl>         <dbl>         <dbl>
## 1 Lisbon         0.08         0.1         0.12         0.05
## 2 Oporto         0.08         0.08         0.11         0.06
## 3 Other regio~    0.11         0.1         0.11         0.04
## # i 2 more variables: Detergents_Paper_mean_norm <dbl>,
## #   Delicatessen_mean_norm <dbl>
```

```
# Non-normalized results
```

```
# All groupings
```

```
wholesale_groups = wholesale %>% group_by(Channel,Region) %>% summarise(Fresh_mean = mean(Fresh),Milk_mean = mean(Milk_mean),
wholesale_groups %>% mutate_if(is.numeric, round, digits=2)
```

```
## # A tibble: 6 x 8
##   Channel Region      Fresh_mean Milk_mean Grocery_mean Frozen_mean
##   <chr>   <chr>         <dbl>     <dbl>         <dbl>         <dbl>
## 1 Horeca Lisbon      12902.    3870.         4026.         3127.
## 2 Horeca Oporto      11651.    2304.         4396.         5745.
## 3 Horeca Other regions 13878.    3487.         3887.         3657.
## 4 Retail Lisbon       5200     10784         18472.         2584.
## 5 Retail Oporto       7290.     9191.         16326.         1541.
## 6 Retail Other regions 9832.    10981.         15954.         1513.
## # i 2 more variables: Detergents_Paper_mean <dbl>, Delicatessen_mean <dbl>
```

```
# Group by Channel
```

```
wholesale_Channel = wholesale_groups %>% group_by(Channel) %>% summarise(Fresh_mean = mean(Fresh_mean), Milk_mean = mean(Milk_mean),
wholesale_Channel %>% mutate_if(is.numeric, round, digits=2)
```

```
## # A tibble: 2 x 7
##   Channel Fresh_mean_norm Milk_mean_norm Grocery_mean_norm Frozen_mean_norm Detergents_Paper_mean_norm
##   <chr>         <dbl>         <dbl>         <dbl>         <dbl>         <dbl>
## 1 Horeca      12810.      3220.         4103.         4176.         740.
## 2 Retail      7440.      10319.        16917.        1879.        7845.
## # i 1 more variable: Delicatessen_mean <dbl>
```

```
# Group by Region
```

```
wholesale_region = wholesale_groups %>% group_by(Region) %>% summarise(Fresh_mean = mean(Fresh_mean), Milk_mean = mean(Milk_mean),
wholesale_region %>% mutate_if(is.numeric, round, digits=2)
```

```
## # A tibble: 3 x 7
##   Region      Fresh_mean Milk_mean Grocery_mean Frozen_mean Detergents_Paper_mean
##   <chr>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
## 1 Lisbon      9051.      7327.      11249.      2856.      4588.
## 2 Oporto      9470.      5748.      10361.      3643.      4446.
## 3 Other reg~ 11855.      7234       9920.      2585.      3843.
## # i 1 more variable: Delicatessen_mean <dbl>
```

Analyse spending patterns across different clusters using visualisations! CLUSTER 1 summary

```
## Cluster 1
```

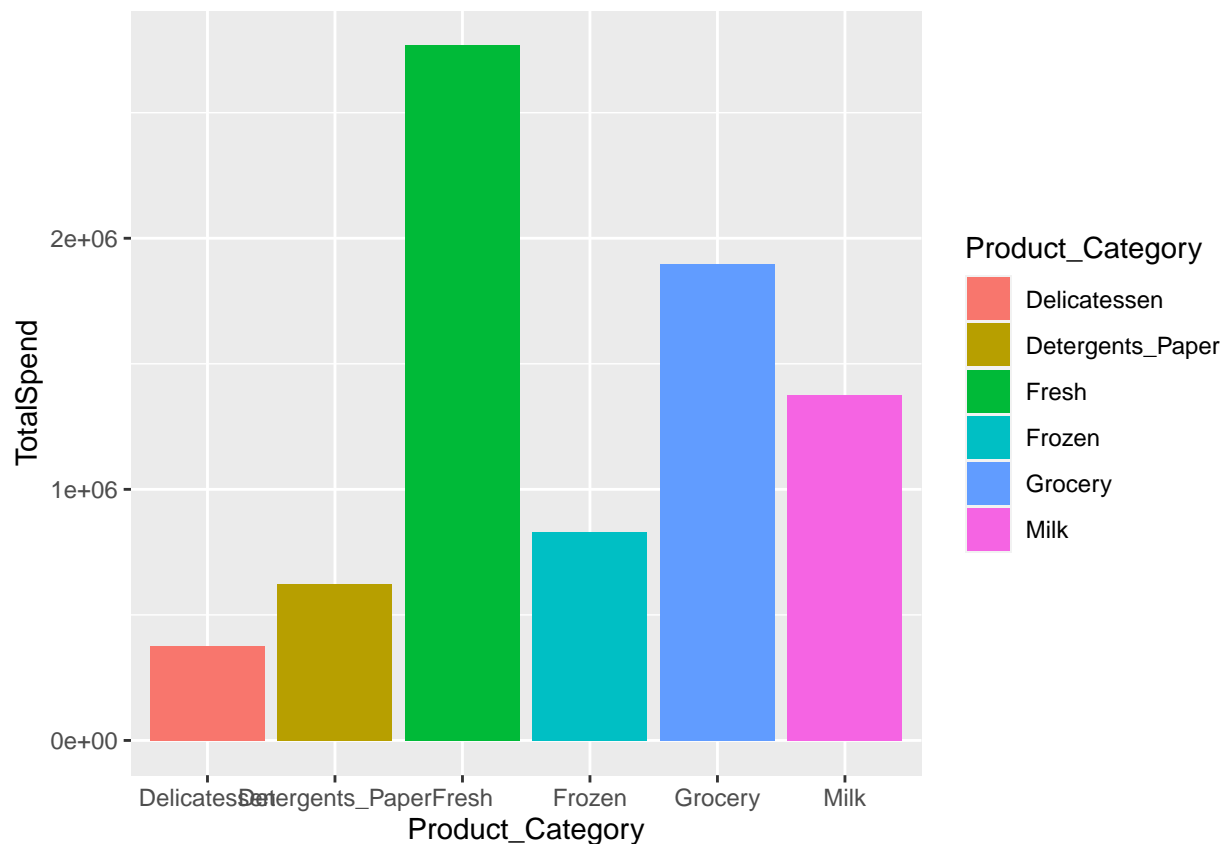
```
cluster1 <- wholesale_pivot[ which(wholesale_pivot$Cluster==1), ]
cluster1 <- subset(cluster1, select = -c(Cluster))
```

```
cluster1_summary <- cluster1 %>% group_by(Channel, Region, Product_Category) %>%
  summarise(TotalSpend = sum(Spend), AvgSpend = mean(Spend))
```

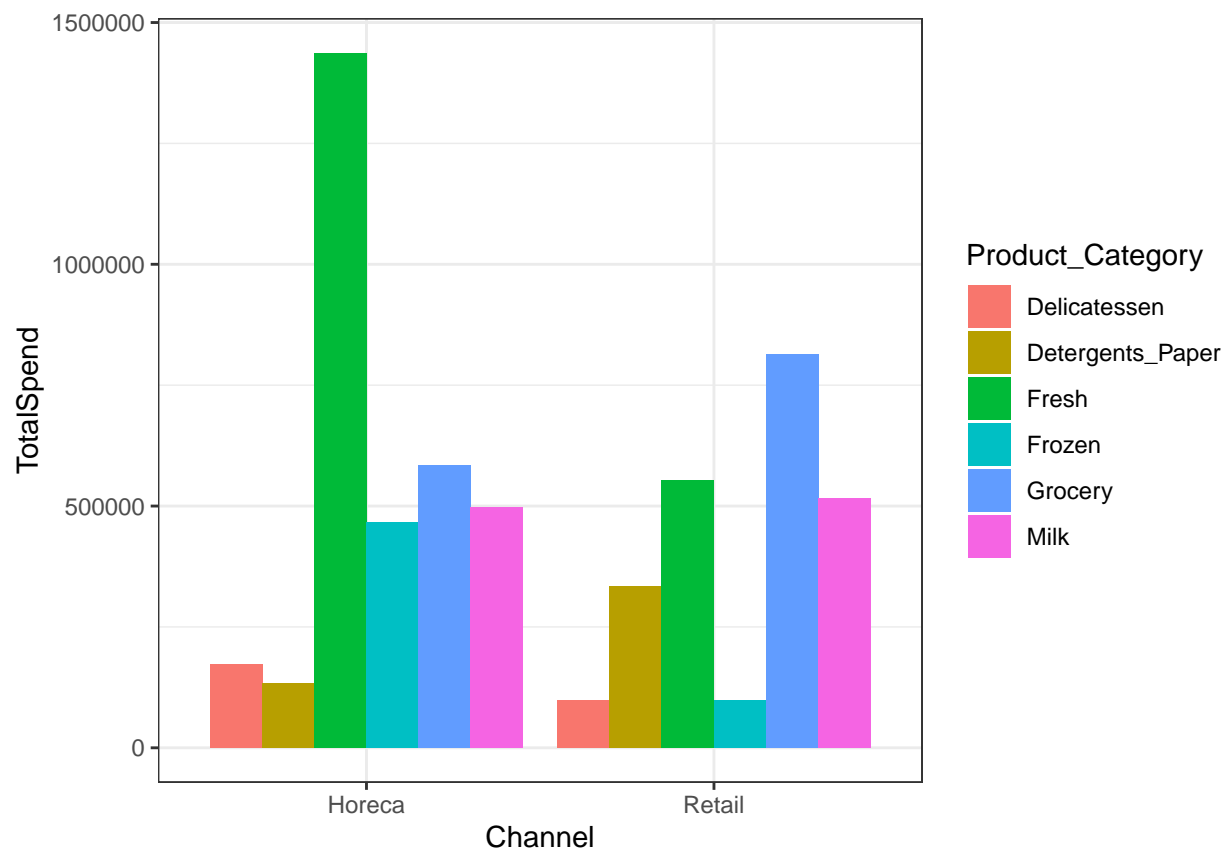
```
## 'summarise()' has grouped output by 'Channel', 'Region'. You can override using
## the '.groups' argument.
```

```
# Plot 1 (Total Spend across categories)
```

```
cluster1_summary %>% ggplot(aes(x = Product_Category, y = TotalSpend, fill = Product_Category)) + geom_bar()
```

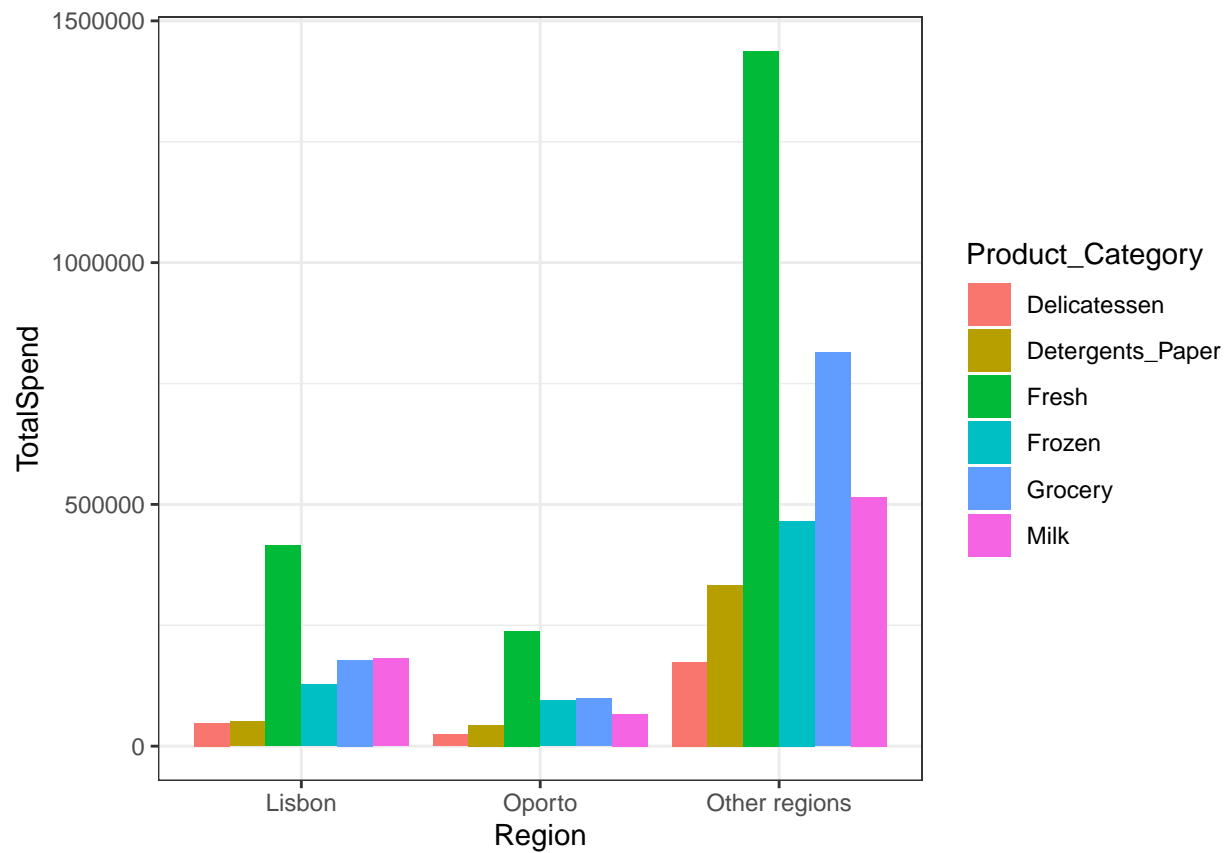


```
# Plot 2 (Total Spend across channels and categories)
cluster1_summary %>% ggplot(aes(Channel, TotalSpend, fill = Product_Category)) + geom_bar(stat = 'ident
```

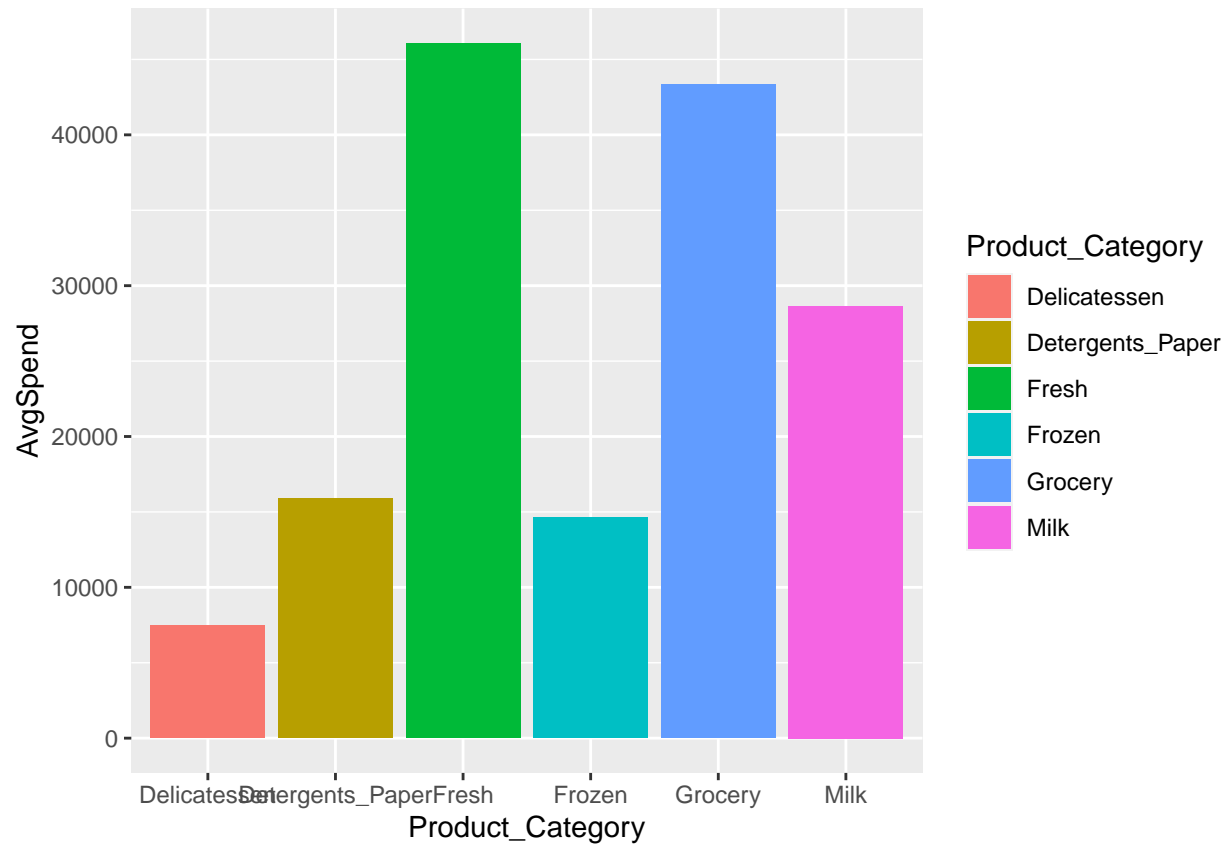


```
# Plot 3 (Total Spend across regions and categories)
cluster1_summary %>% ggplot(aes(Region, TotalSpend, fill = Product_Category)) + geom_bar(stat = 'ident
```

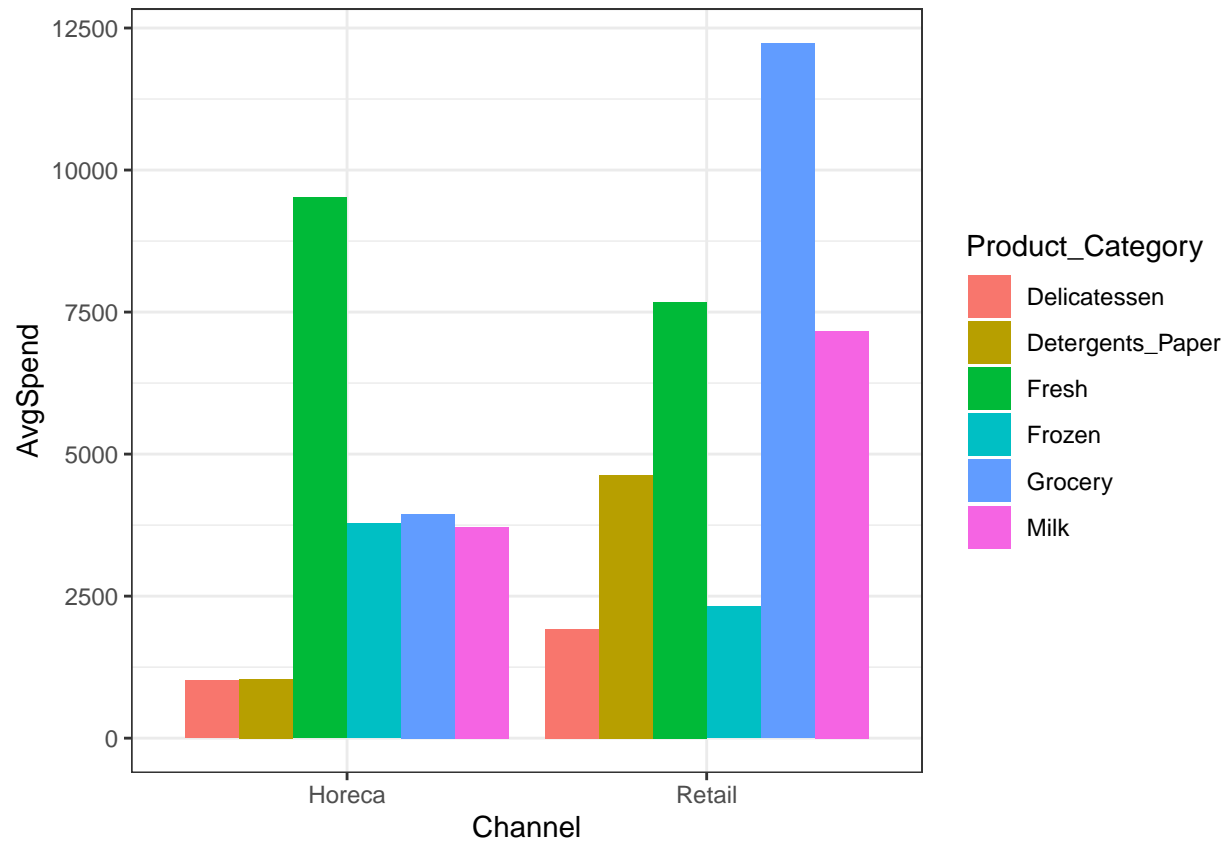




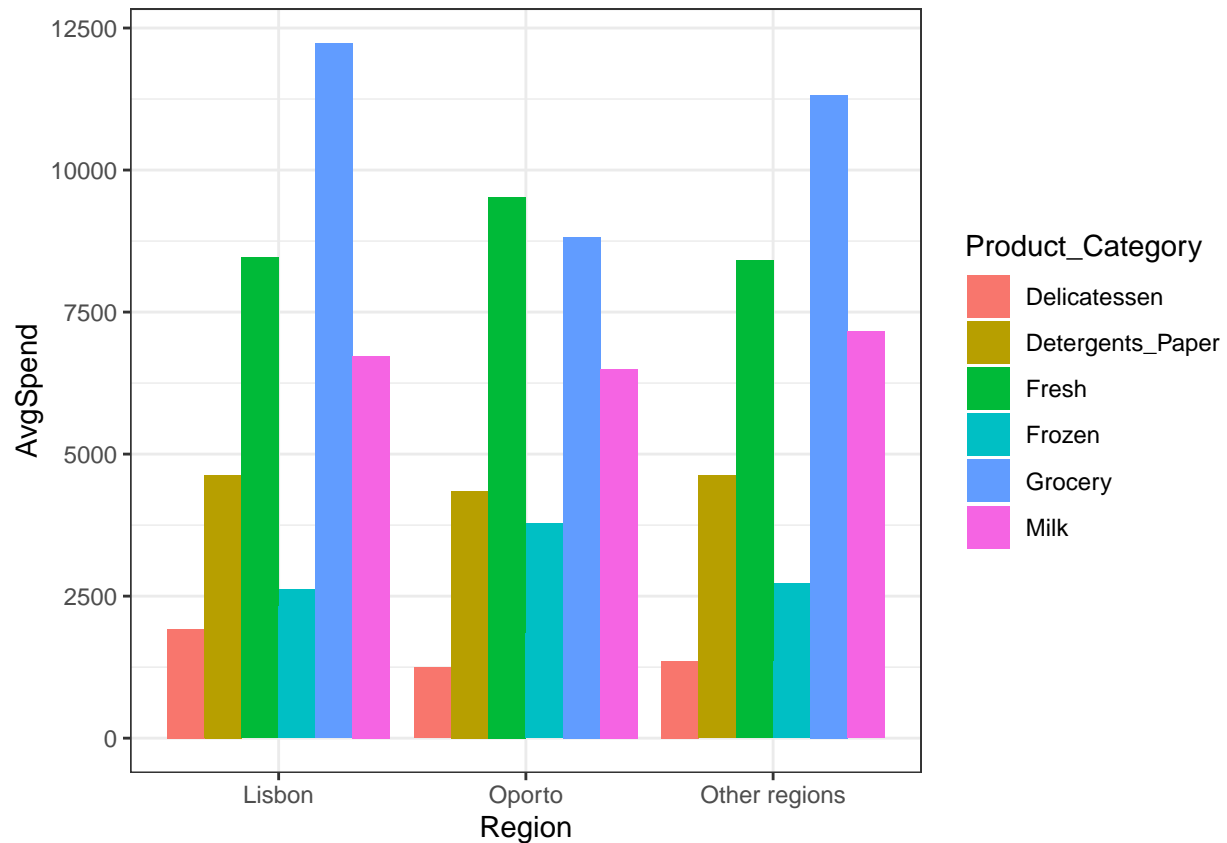
```
# Plot 4 (Average Spend across categories)
cluster1_summary %>% ggplot(aes(x = Product_Category, y = AvgSpend, fill = Product_Category)) + geom_col()
```



```
# Plot 5 (Average Spend across channels and categories)
cluster1_summary %>% ggplot(aes(Channel, AvgSpend, fill = Product_Category)) + geom_bar(stat = 'identity')
```



```
# Plot 6 (Average Spend across regions and categories)
cluster1_summary %>% ggplot(aes(Channel, AvgSpend, fill = Product_Category)) + geom_bar(stat = 'identity')
```



CLUSTER 2 summary

```
## Cluster 2
```

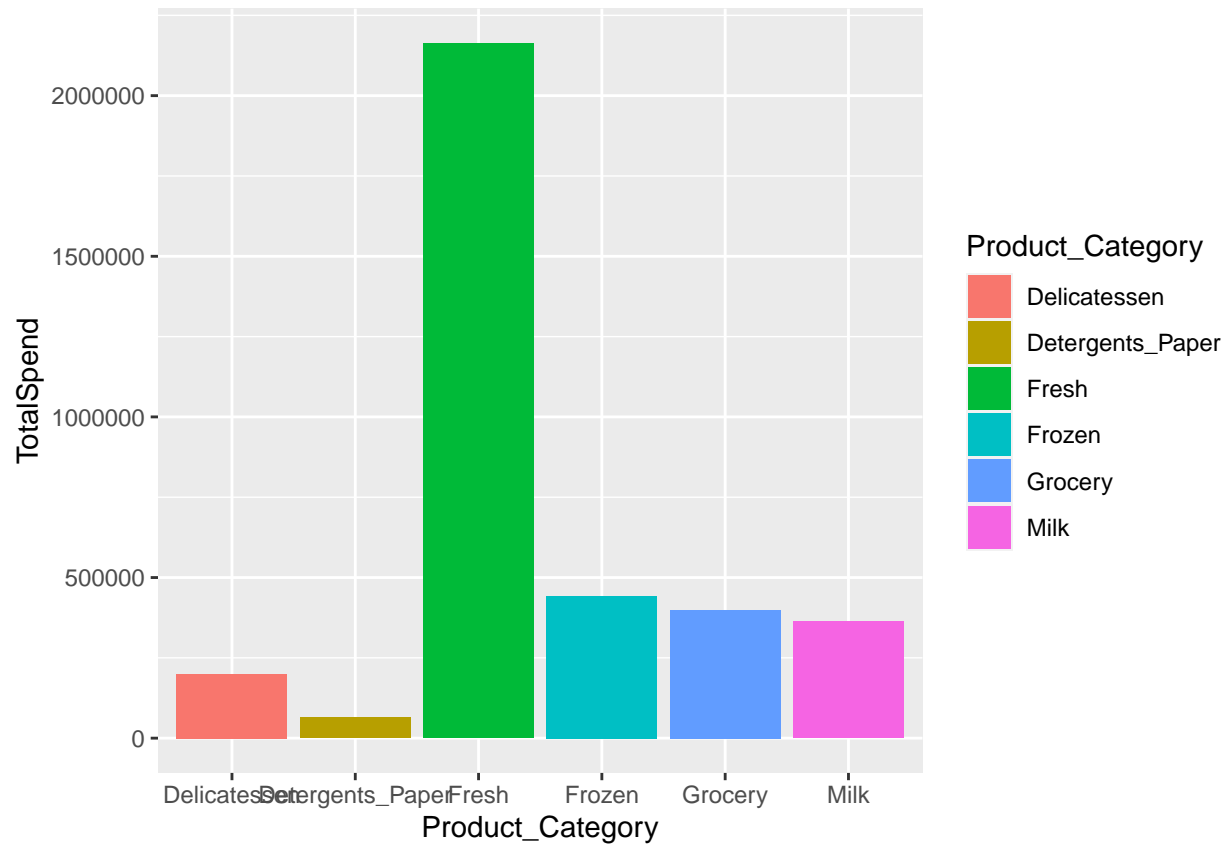
```
cluster2 <- wholesale_pivot[ which(wholesale_pivot$Cluster==2), ]
cluster2 <- subset(cluster2, select = -c(Cluster))
```

```
cluster2_summary <- cluster2 %>% group_by(Channel, Region, Product_Category) %>%
  summarise(TotalSpend = sum(Spend), AvgSpend = mean(Spend))
```

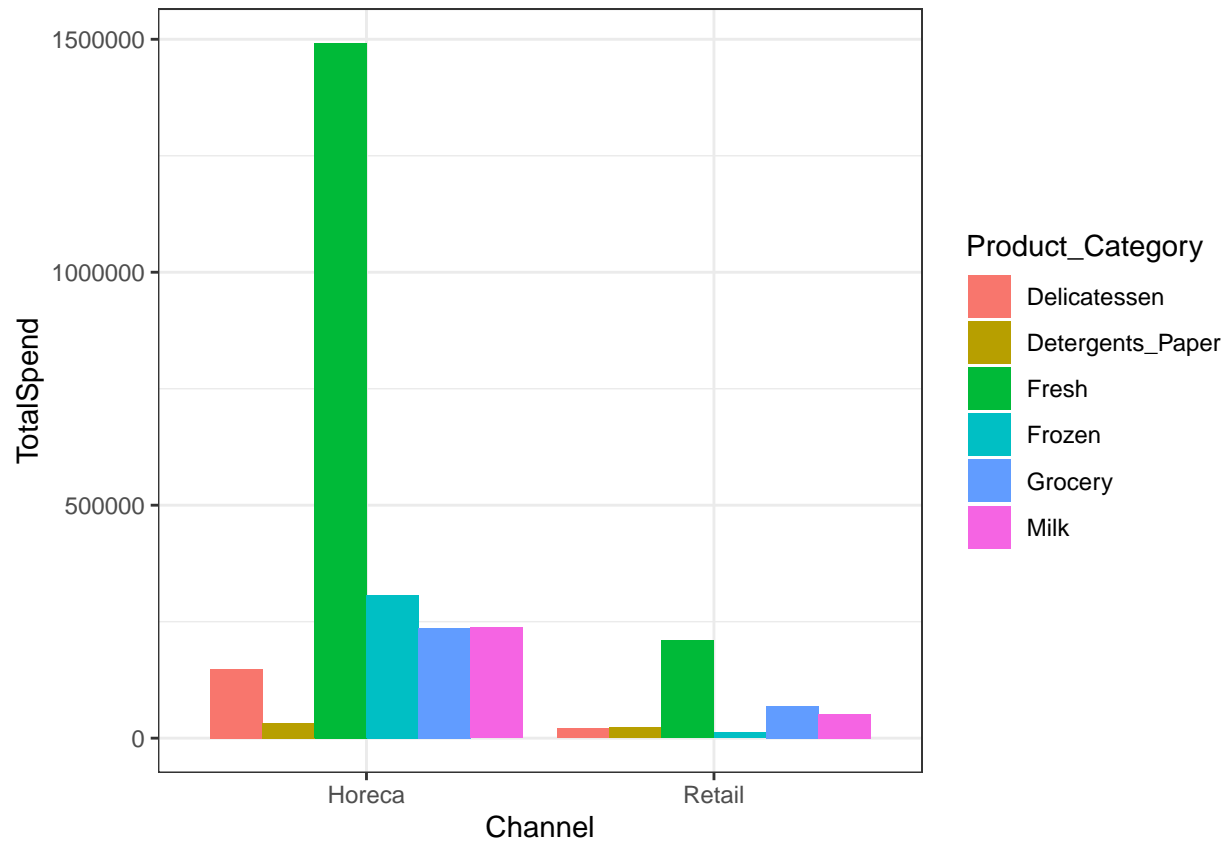
```
## 'summarise()' has grouped output by 'Channel', 'Region'. You can override using
## the '.groups' argument.
```

```
# Plot 1 (Total Spend across categories)
```

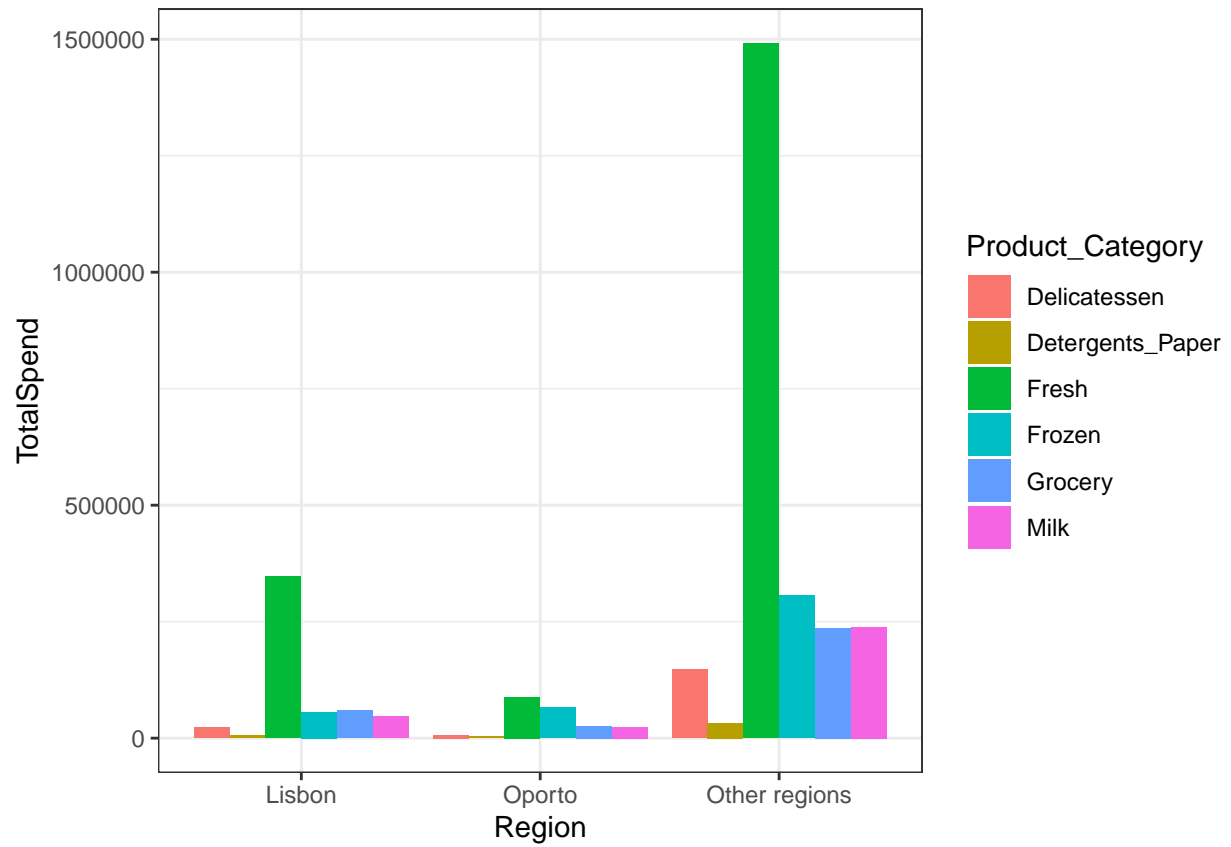
```
cluster2_summary %>% ggplot(aes(x = Product_Category, y = TotalSpend, fill = Product_Category)) + geom_bar()
```



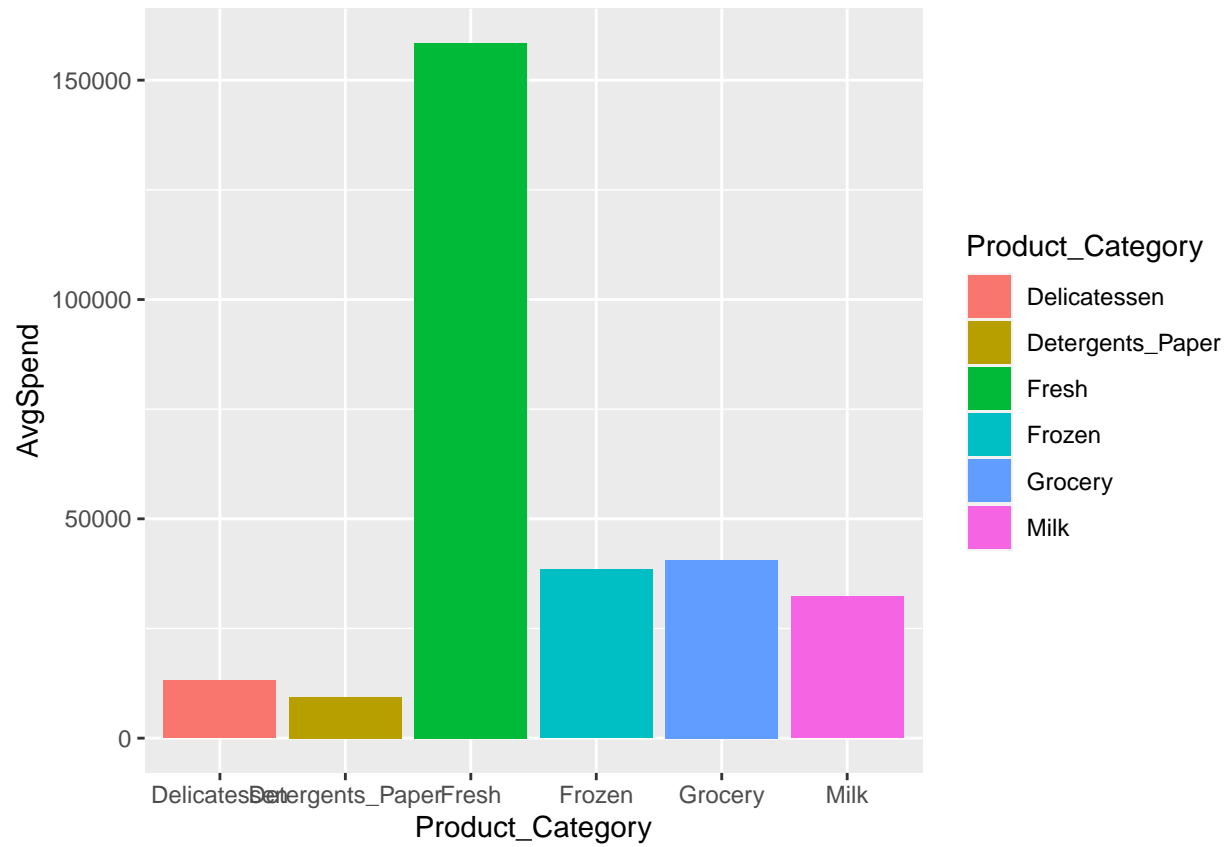
```
# Plot 2 (Total Spend across channels and categories)
cluster2_summary %>% ggplot(aes(Channel, TotalSpend, fill = Product_Category)) + geom_bar(stat = 'ident
```



```
# Plot 3 (Total Spend across regions and categories)
cluster2_summary %>% ggplot(aes(Channel, TotalSpend, fill = Product_Category)) + geom_bar(stat = 'identity')
```

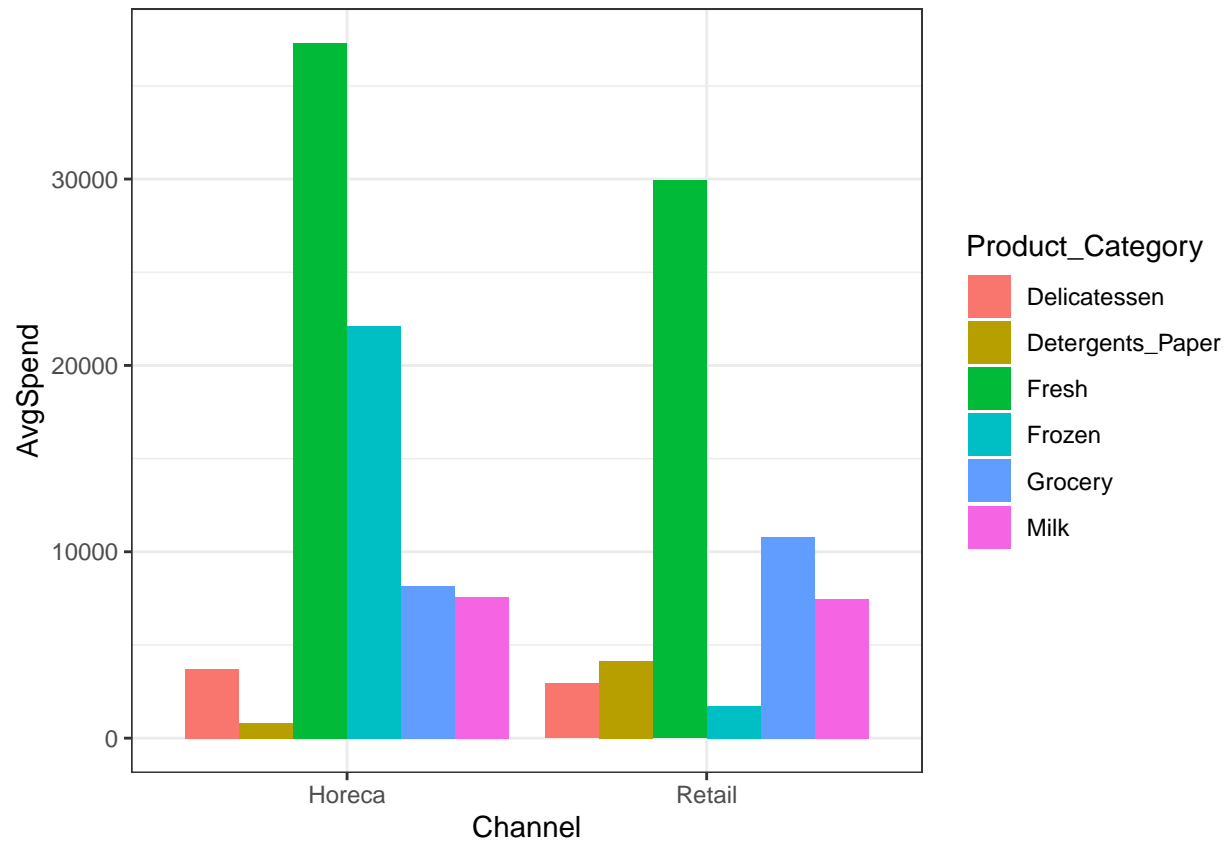


```
# Plot 4 (Average Spend across categories)
cluster2_summary %>% ggplot(aes(x = Product_Category, y = AvgSpend, fill = Product_Category)) + geom_col()
```

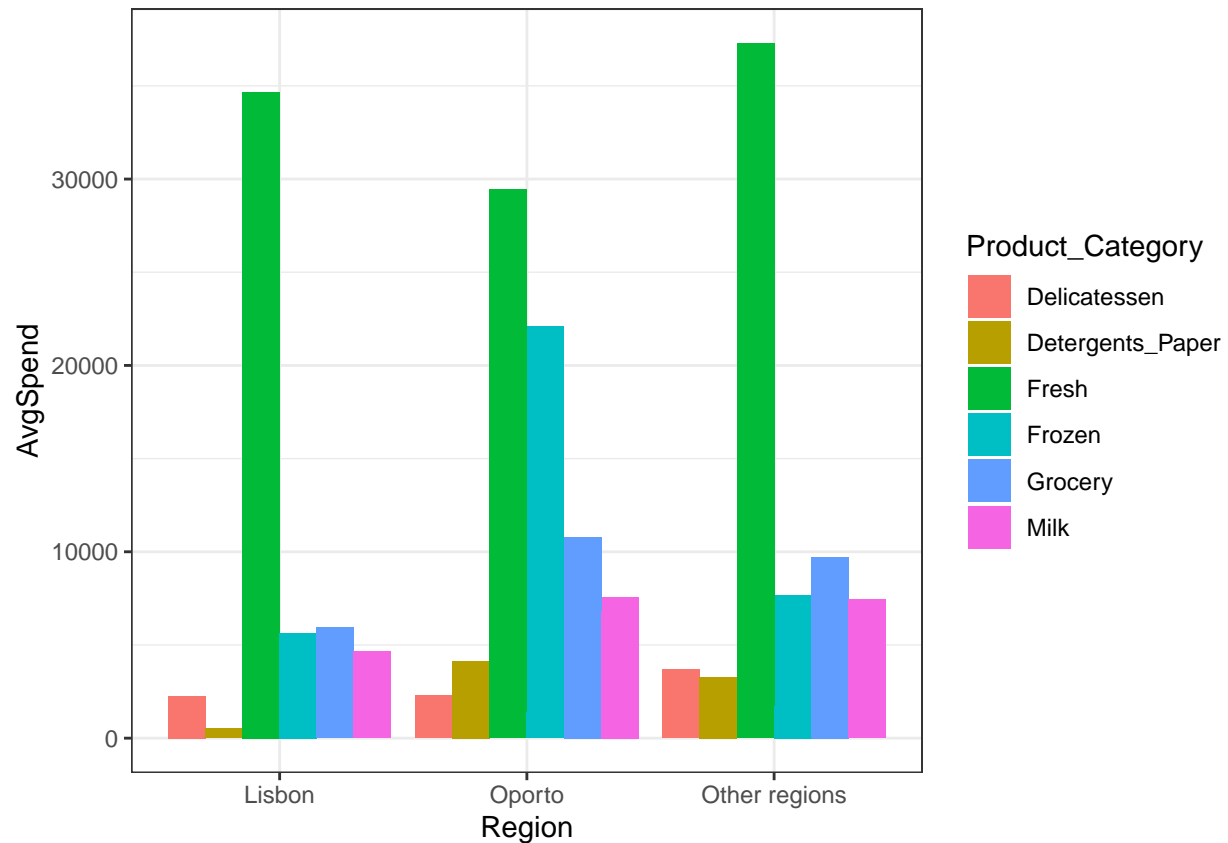


```
# Plot 5 (Average Spend across channels and categories)
cluster2_summary %>% ggplot(aes(Channel, AvgSpend, fill = Product_Category)) + geom_bar(stat = 'identity')
```





```
# Plot 6 (Average Spend across regions and categories)
cluster2_summary %>% ggplot(aes(Channel, AvgSpend, fill = Product_Category)) + geom_bar(stat = 'identity')
```



CLUSTER 3 summary

*## Cluster 3*

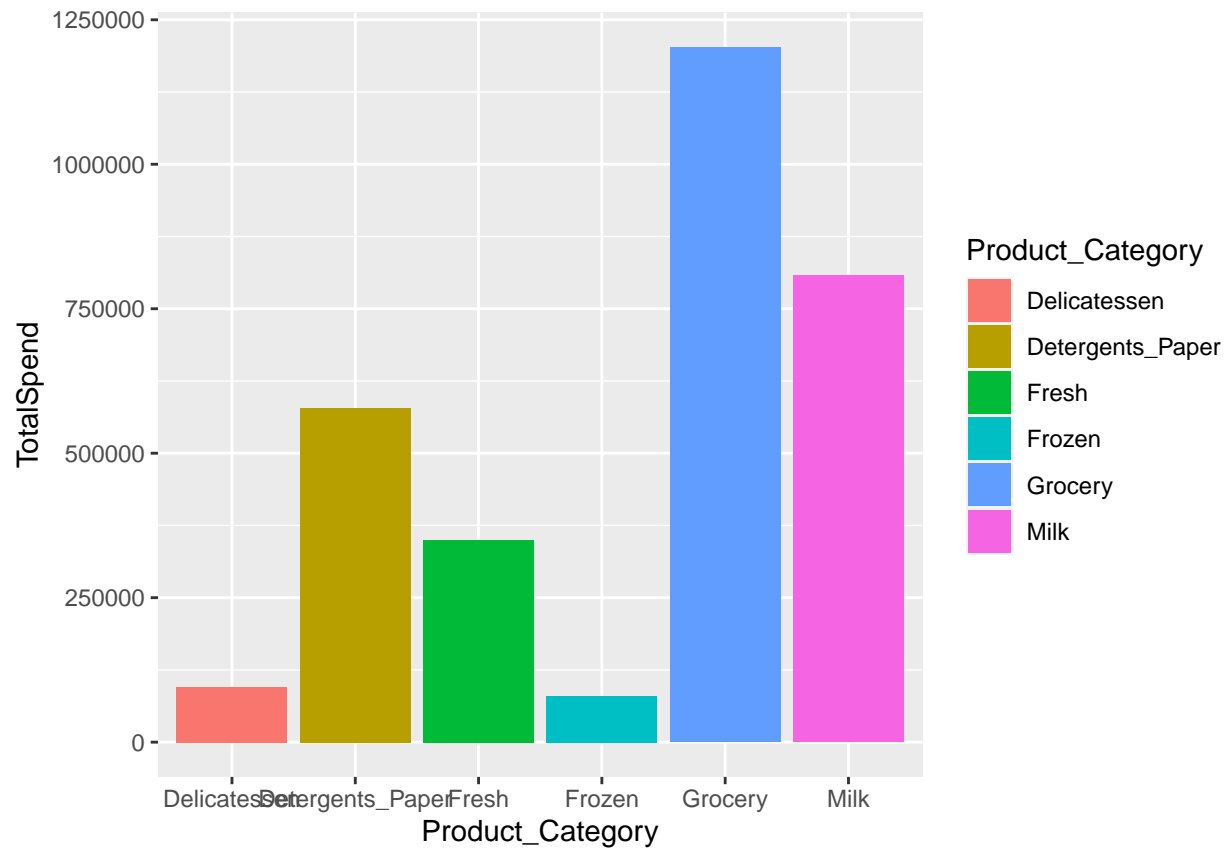
```
cluster3 <- wholesale_pivot[ which(wholesale_pivot$Cluster==3), ]
cluster3 <- subset(cluster3, select = -c(Cluster))
```

```
cluster3_summary <- cluster3 %>% group_by(Channel, Region, Product_Category) %>%
  summarise(TotalSpend = sum(Spend), AvgSpend = mean(Spend))
```

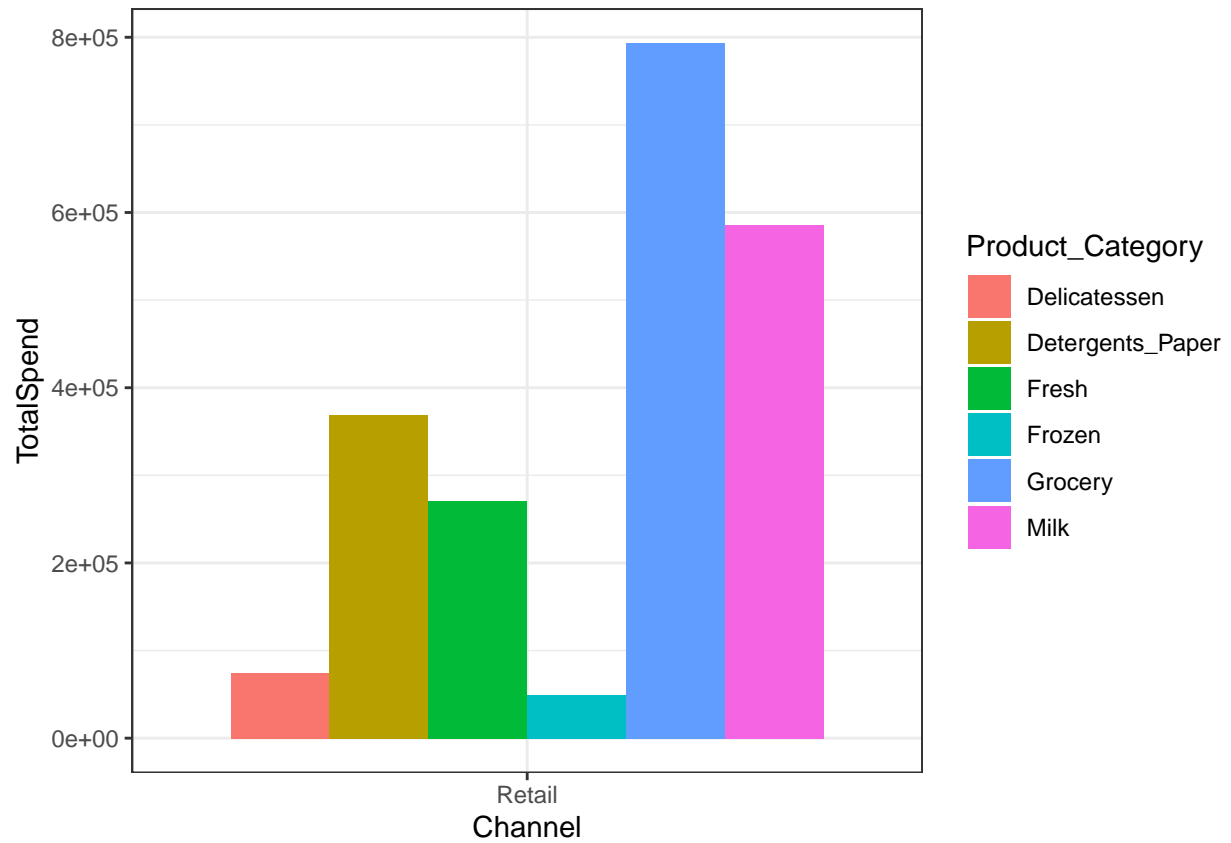
## 'summarise()' has grouped output by 'Channel', 'Region'. You can override using  
## the '.groups' argument.

*# Plot 1 (Total Spend across categories)*

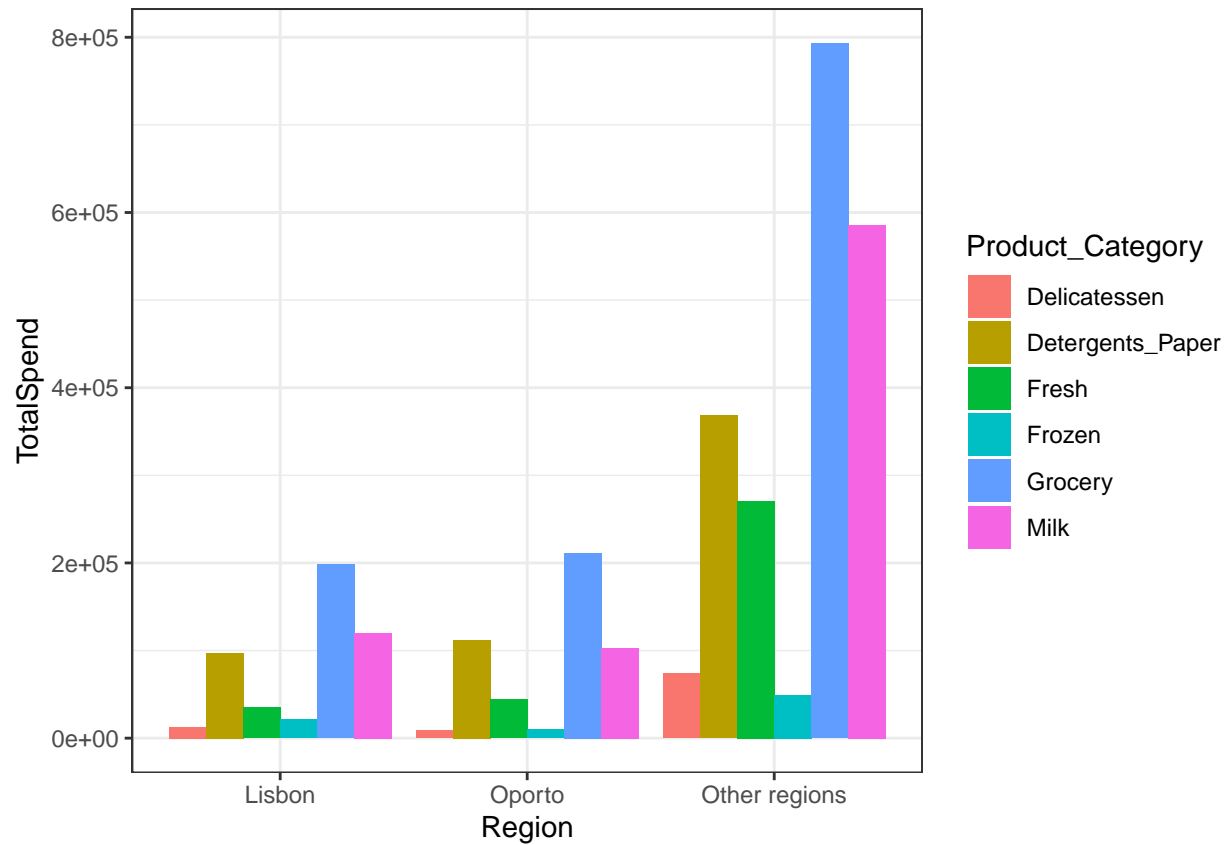
```
cluster3_summary %>% ggplot(aes(x = Product_Category, y = TotalSpend, fill = Product_Category)) + geom_bar()
```



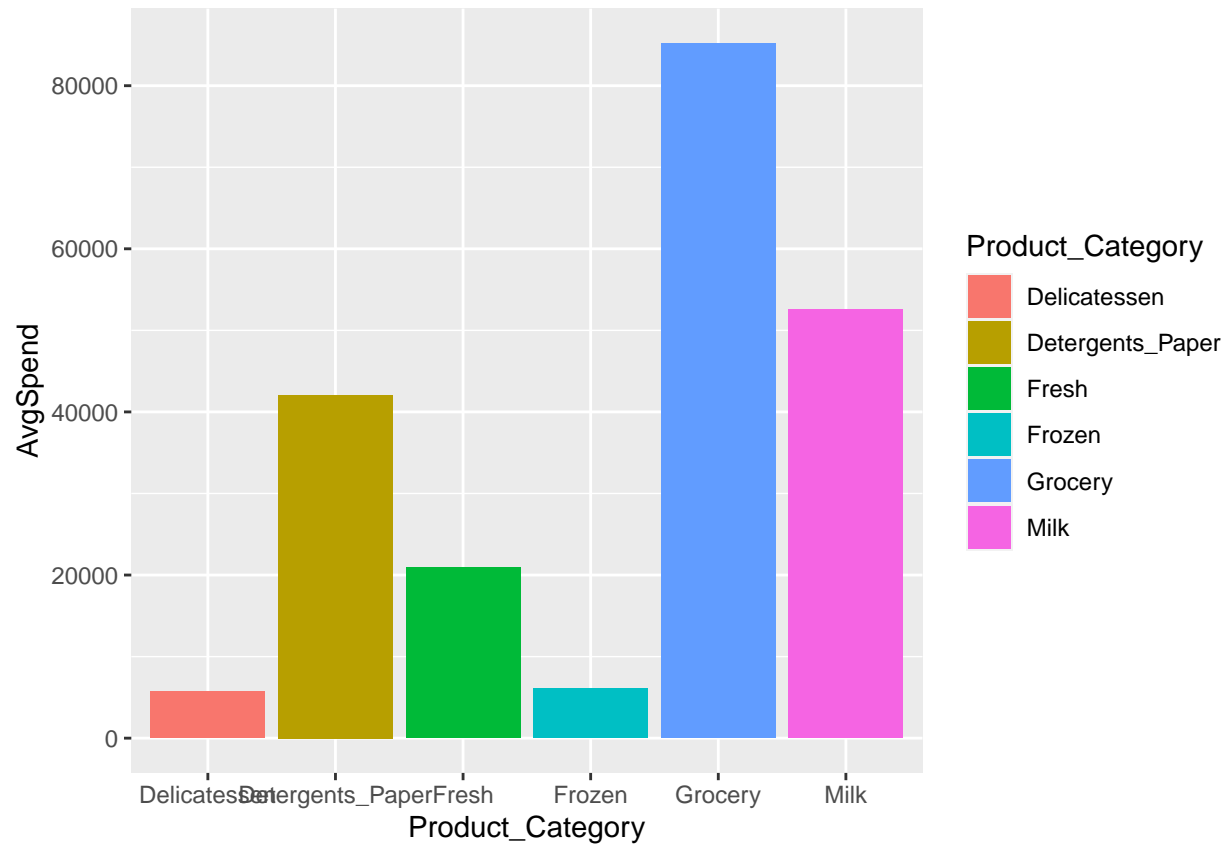
```
# Plot 2 (Total Spend across channels and categories)
cluster3_summary %>% ggplot(aes(Channel, TotalSpend, fill = Product_Category)) + geom_bar(stat = 'ident
```



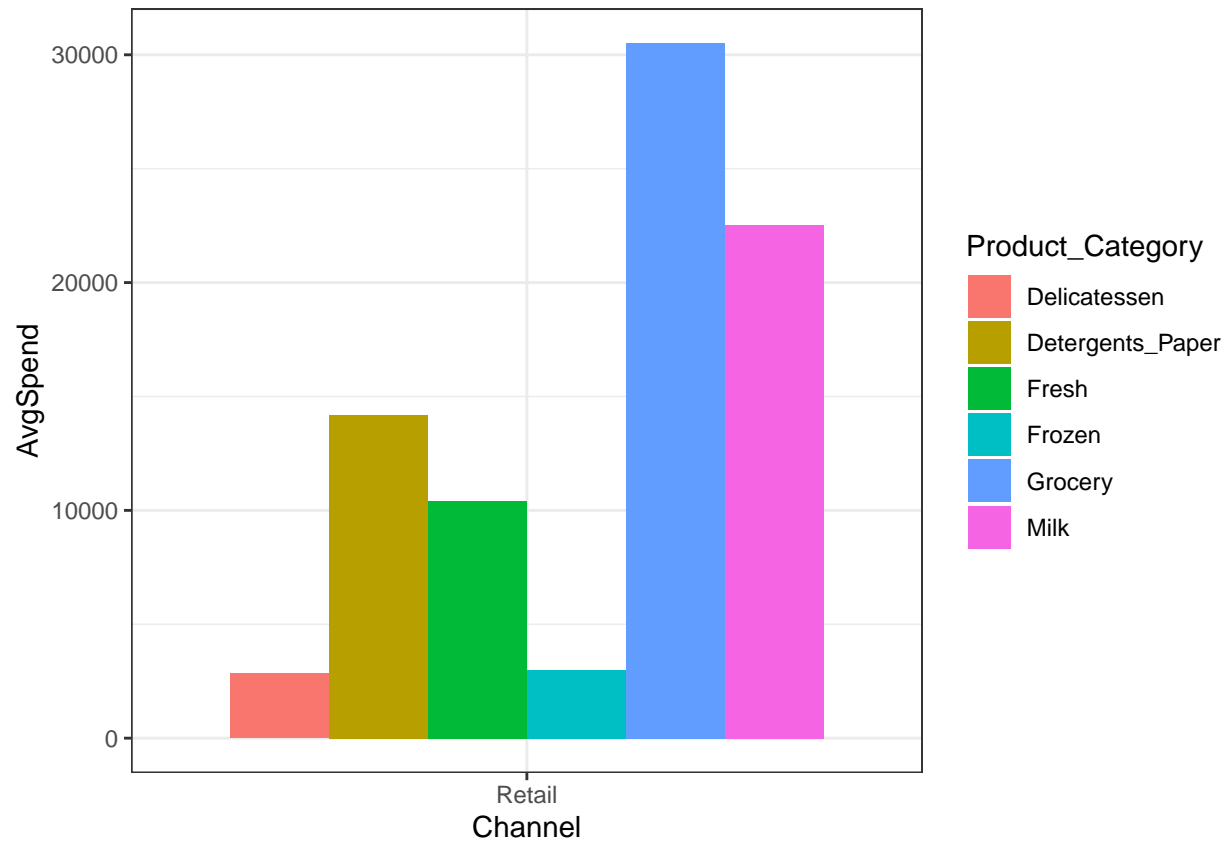
```
# Plot 3 (Total Spend across regions and categories)
cluster3_summary %>% ggplot(aes(Retail, TotalSpend, fill = Product_Category)) + geom_bar(stat = 'identity')
```



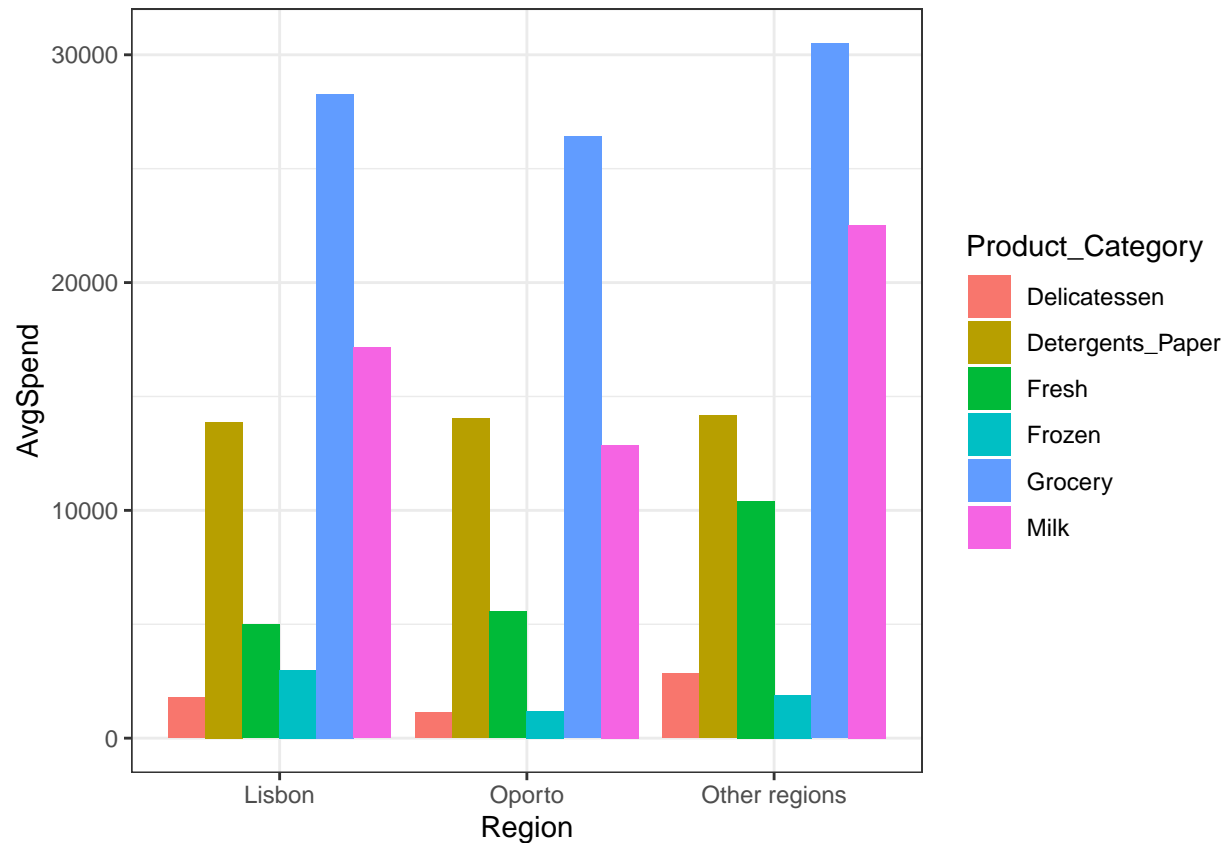
```
# Plot 4 (Average Spend across categories)
cluster3_summary %>% ggplot(aes(x = Product_Category, y = AvgSpend, fill = Product_Category)) + geom_col()
```



```
# Plot 5 (Average Spend across channels and categories)
cluster3_summary %>% ggplot(aes(Channel, AvgSpend, fill = Product_Category)) + geom_bar(stat = 'identity')
```



```
# Plot 6 (Average Spend across regions and categories)
cluster3_summary %>% ggplot(aes(Retail, AvgSpend, fill = Product_Category)) + geom_bar(stat = 'identity')
```



Then we looked into the distribution of clusters across channels and regions. ## We can change the tables into better visualizations

```
# Composition counts for each cluster
cluster_composition = table(wholesale_clustered$Channel, wholesale_clustered$Region, wholesale_clustered$Product_Category)
cluster_composition
```

```
## , , = 1
##
##
##      Lisbon Oporto Other regions
## Horeca    49    25         171
## Retail    11    10          72
##
## , , = 2
##
##
##      Lisbon Oporto Other regions
## Horeca    10     3          40
## Retail     0     1           7
##
## , , = 3
##
##
##      Lisbon Oporto Other regions
## Horeca     0     0           0
```



```
##   Retail      7      8      26
```

```
percentages = cluster_composition
# Composition percentages for each cluster
for (i in 1:k){
  total = sum(percentages[,i])
  percentages[,i] = t(round(t(percentages[,i])/total,2))
}
percentages
```

```
## , , = 1
##
##
##           Lisbon Oporto Other regions
##   Horeca   0.14   0.07           0.51
##   Retail   0.03   0.03           0.21
##
## , , = 2
##
##
##           Lisbon Oporto Other regions
##   Horeca   0.16   0.05           0.66
##   Retail   0.00   0.02           0.11
##
## , , = 3
##
##
##           Lisbon Oporto Other regions
##   Horeca   0.00   0.00           0.00
##   Retail   0.17   0.20           0.63
```

The data suggests the following:

CLUSTER 1 - This cluster constitutes of the clients majorly from Horeca sector (i.e. Hotels/Restaurants/Cafes) approx. 72%. These clients demonstrate high spending in the Fresh category, while the remaining 28% of the clients from Retail channel spend more on Grocery. Clients from both channels put together have spends across these 3 categories majorly - Fresh, Grocery and Milk.

CLUSTER 2 - This cluster constitutes of the clients only from Retail channel, majorly from the Other regions (63%). Their spending pattern shows huge spends in the Grocery, followed by Milk and Detergents & Paper.

CLUSTER 3 - This cluster constitutes of clients from both Horeca and Retail, majorly from Other regions (66%). These clients spend mostly in the Fresh category. The spend across other categories is minimal.