

# DAR F23 Project Status Notebook - Assignment 06

## Hockey Analytics

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### Weekly Work Summary

- RCS ID: smithc22
- Project Name: Hockey Analytics
- Summary of work since last week

I analyzed Jeff's discretization of clusters, using UMAP + Kmeans as well as PCA to see if the categorization of everything would fix PCA's issue of not treating continuous and categorical variables evenly.

I also added a slider to allow the user to subset the data based on features in Play\_Plotter. I'll discuss at the next meeting whether to allow the user to slice data on multiple features and how to implement it in a manner that would make sense in the integration

Finally, I attended the CommD meeting with Ashley about colors to get some advice for the color scheme. The new color scheme for clusters has been applied. They also recommend we switch the rink plot to grayscale.

- Summary of github issues added and worked

n/a

- Summary of github commits

Branch Name: dar-smithc22 Files pushed: [https://github.rpi.edu/DataINCITE/Hockey\\_Fall\\_2023/blob/dar-smithc22/ShinyApps/Play\\_Plotter/app.R](https://github.rpi.edu/DataINCITE/Hockey_Fall_2023/blob/dar-smithc22/ShinyApps/Play_Plotter/app.R)

[https://github.rpi.edu/DataINCITE/Hockey\\_Fall\\_2023/blob/dar-smithc22/StudentNotebooks/Assignment06/smithc22\\_assignment06.Rmd](https://github.rpi.edu/DataINCITE/Hockey_Fall_2023/blob/dar-smithc22/StudentNotebooks/Assignment06/smithc22_assignment06.Rmd)

[https://github.rpi.edu/DataINCITE/Hockey\\_Fall\\_2023/raw/dar-smithc22/StudentNotebooks/Assignment06/smithc22\\_assignment06.pdf](https://github.rpi.edu/DataINCITE/Hockey_Fall_2023/raw/dar-smithc22/StudentNotebooks/Assignment06/smithc22_assignment06.pdf)

App Links:

[https://lp01.idea.rpi.edu/shiny/smithc22/Play\\_Plotter/](https://lp01.idea.rpi.edu/shiny/smithc22/Play_Plotter/)

- List of presentations, papers, or other outputs
  - N/A
- List of references (if necessary)

N/A

- Indicate any use of group shared code base

I used some of Lieben's code from his app to help with the sliders in my app

- Indicate which parts of your described work were done by you or as part of joint efforts

Most of the work not already attributed was done by me, choosing colors was a joint effort between me, Ashley, and Jessica from CommD

## Personal Contribution

- Clearly defined, unique contribution(s) done by you: code, ideas, writing...

The code in this notebook is my own. I fixed some bugs in my app relating to cluster colors and also added a method of filtering out features based on a range, although I got the code for the UI from Lieben.

- Include github issues you've addressed

N/A

## Analysis: Question 1: Does categorizing continuous variables change the clustering?

### Question being asked

*Provide in natural language a statement of what question you're trying to answer*

How does categorizing all the data change the clustering performed by UMAP?

### Data Preparation

*Provide in natural language a description of the data you are using for this analysis*

The categorical data from Jeff, where each continuous feature is put into tertiles.

*Include a step-by-step description of how you prepare your data for analysis*

1. Read in Jeff's data
2. Select only the categorized parts
3. Set things up so UMAP works properly *If you're re-using dataframes prepared in another section, simply re-state what data you're using*

```
# Include all data processing code (if necessary), clearly commented
library(data.table)
library(mltools)
library(dplyr)
```

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

## The following objects are masked from 'package:data.table':
##
##   between, first, last
```

```

## The following objects are masked from 'package:stats':
##
##   filter, lag
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
library(plotly)

## Loading required package: ggplot2
##
## Attaching package: 'plotly'
## The following object is masked from 'package:ggplot2':
##
##   last_plot
## The following object is masked from 'package:stats':
##
##   filter
## The following object is masked from 'package:graphics':
##
##   layout
library(umap)
library(ggplot2)
library(scatterplot3d)
library(rgl)

## Warning in rgl.init(initValue, onlyNULL): RGL: unable to open X11 display
## Warning: 'rgl.init' failed, running with 'rgl.useNULL = TRUE'.
library(car)

## Loading required package: carData
##
## Attaching package: 'car'
## The following object is masked from 'package:dplyr':
##
##   recode
library(grid)
library(gridExtra)

##
## Attaching package: 'gridExtra'
## The following object is masked from 'package:dplyr':
##
##   combine
library(kableExtra)

##
## Attaching package: 'kableExtra'

```

```

## The following object is masked from 'package:dplyr':
##
##   group_rows
library(heatmaply)

## Loading required package: viridis
## Loading required package: viridisLite
##
## =====
## Welcome to heatmaply version 1.4.2
##
## Type citation('heatmaply') for how to cite the package.
## Type ?heatmaply for the main documentation.
##
## The github page is: https://github.com/talgalili/heatmaply/
## Please submit your suggestions and bug-reports at: https://github.com/talgalili/heatmaply/issues
## You may ask questions at stackoverflow, use the r and heatmaply tags:
##   https://stackoverflow.com/questions/tagged/heatmaply
## =====
library(gplots)

##
## Attaching package: 'gplots'
## The following object is masked from 'package:stats':
##
##   lowess
library(tibble)
library(ggbiplot)

## Loading required package: plyr
## -----
## You have loaded plyr after dplyr - this is likely to cause problems.
## If you need functions from both plyr and dplyr, please load plyr first, then dplyr:
## library(plyr); library(dplyr)
## -----
##
## Attaching package: 'plyr'
## The following objects are masked from 'package:plotly':
##
##   arrange, mutate, rename, summarise
## The following objects are masked from 'package:dplyr':
##
##   arrange, count, desc, failwith, id, mutate, rename, summarise,
##   summarize
## Loading required package: scales
##
## Attaching package: 'scales'

```

```
## The following object is masked from 'package:viridis':
##
##   viridis_pal

library(stats)
setwd("~/Hockey_Fall_2023/StudentNotebooks/Assignment06")
shots_stats.df <- readRDS("~/Hockey_Fall_2023/StudentData/categorized_shots_stats_goal.df.Rds")
justCat <- cbind.data.frame(shots_stats.df[,16:23], shots_stats.df[,8:10])
str(justCat)

## 'data.frame':   210 obs. of  11 variables:
##  $ puckSpeedCategory      : num  2 1 1 1 2 2 2 2 1 2 ...
##  $ puckAngleCategory      : num  0 1 0 2 0 1 0 1 0 1 ...
##  $ puckDistCategory       : num  2 1 1 1 0 2 0 1 2 2 ...
##  $ posTimeCategory        : num  0 0 0 2 2 1 2 2 1 0 ...
##  $ goalieDist_qCategory   : num  2 0 1 0 2 1 1 1 1 0 ...
##  $ shooterSpeed_qCategory : num  1 0 1 1 2 0 2 2 0 0 ...
##  $ goalieAngleCategory    : num  0 1 0 2 1 1 0 1 1 1 ...
##  $ defDistCategory        : num  2 0 2 0 1 1 1 2 1 1 ...
##  $ NumOffense              : int  1 1 0 0 1 2 0 2 2 1 ...
##  $ NumDefense              : int  2 2 2 2 2 3 1 2 4 2 ...
##  $ rightHanded             : num  0 0 0 0 1 0 0 0 1 1 ...

custom.config <- umap.defaults
custom.config$random_state <- 11102023
set.seed(100)

select <- dplyr::select

wssplot <- function(data, nc=15, seed=100){
  wss <- data.frame(cluster=1:nc, quality=c(0))
  for (i in 1:nc){
    set.seed(seed)
    wss[i,2] <- kmeans(data, centers=i)$tot.withinss}
  ggplot(data=wss,aes(x=cluster,y=quality)) +
    geom_line() +
    ggtitle("Quality of k-means by Cluster")
}
```

## Analysis: Methods and results

*Describe in natural language a statement of the analysis you're trying to do*

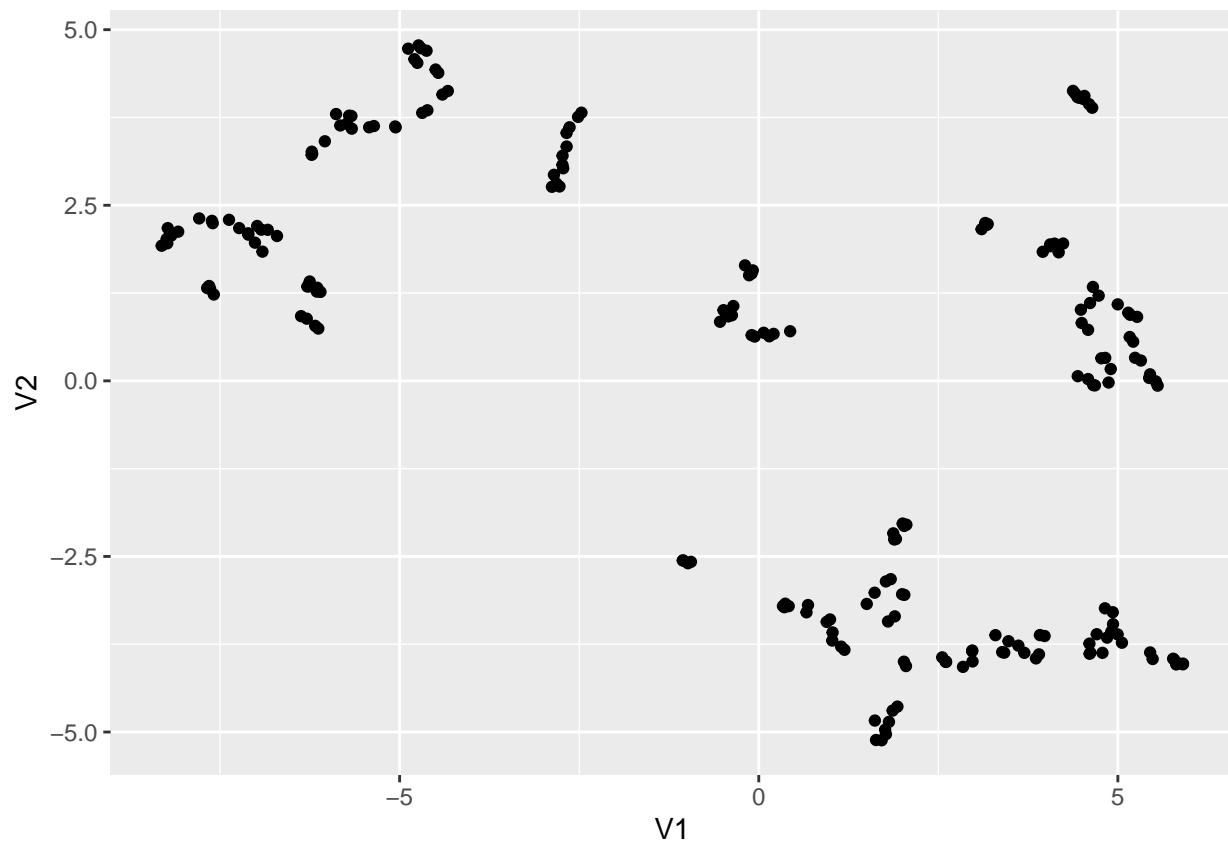
Trying to run UMAP and then Kmeans clustering on Jeff's data

*Provide clearly commented analysis code; include code for tables and figures!*

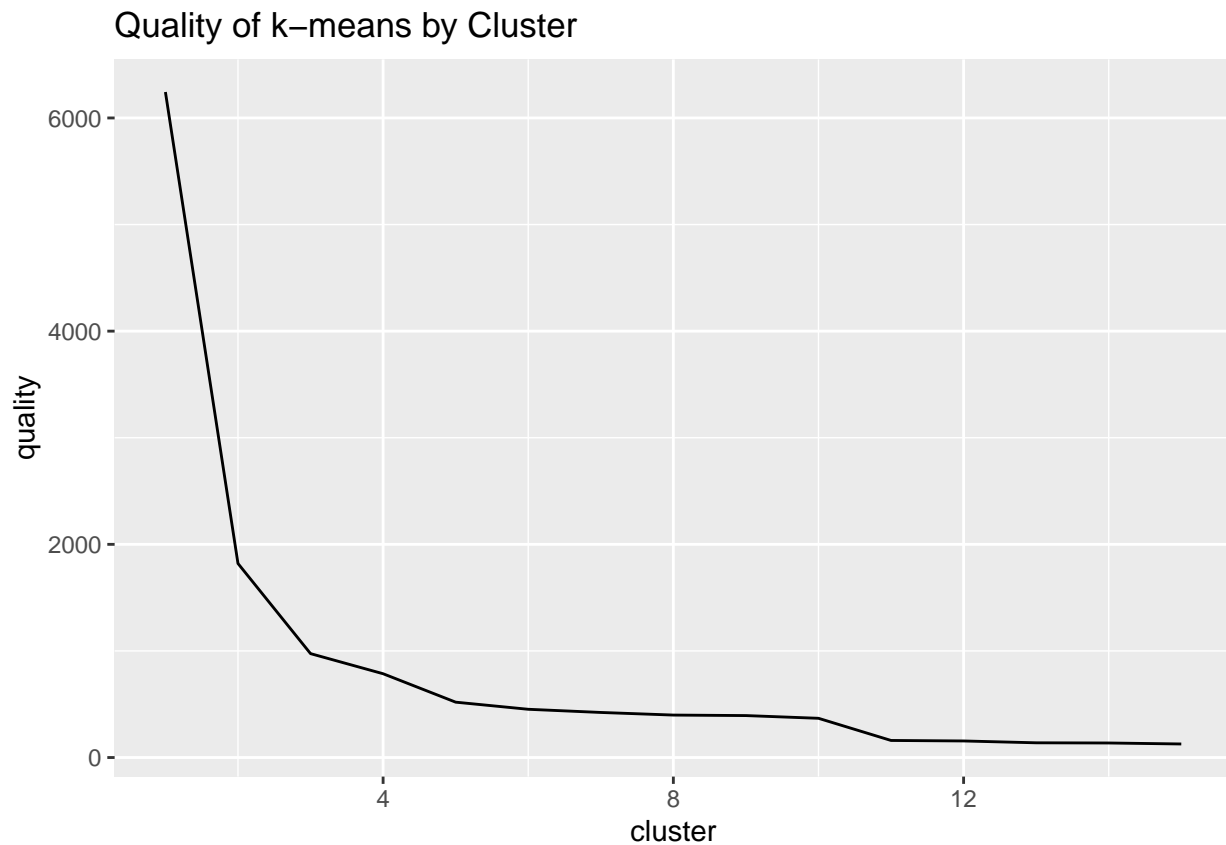
```
# Include all analysis code, clearly commented
# If not possible, screen shots are acceptable.
# If your contributions included things that are not done in an R-notebook,
# (e.g. researching, writing, and coding in Python), you still need to do
# this status notebook in R. Describe what you did here and put any products
# that you created in github. If you are writing online documents (e.g. overleaf
# or google docs), you can include links to the documents in this notebook
# instead of actual text.

umapCat <- umap(justCat,n_components = 2, config = custom.config)
```

```
ggplot()+  
  geom_point(data = as.data.frame(umapCat$layout),aes(x = V1,y = V2))
```



```
wssplot(umapCat$layout)
```



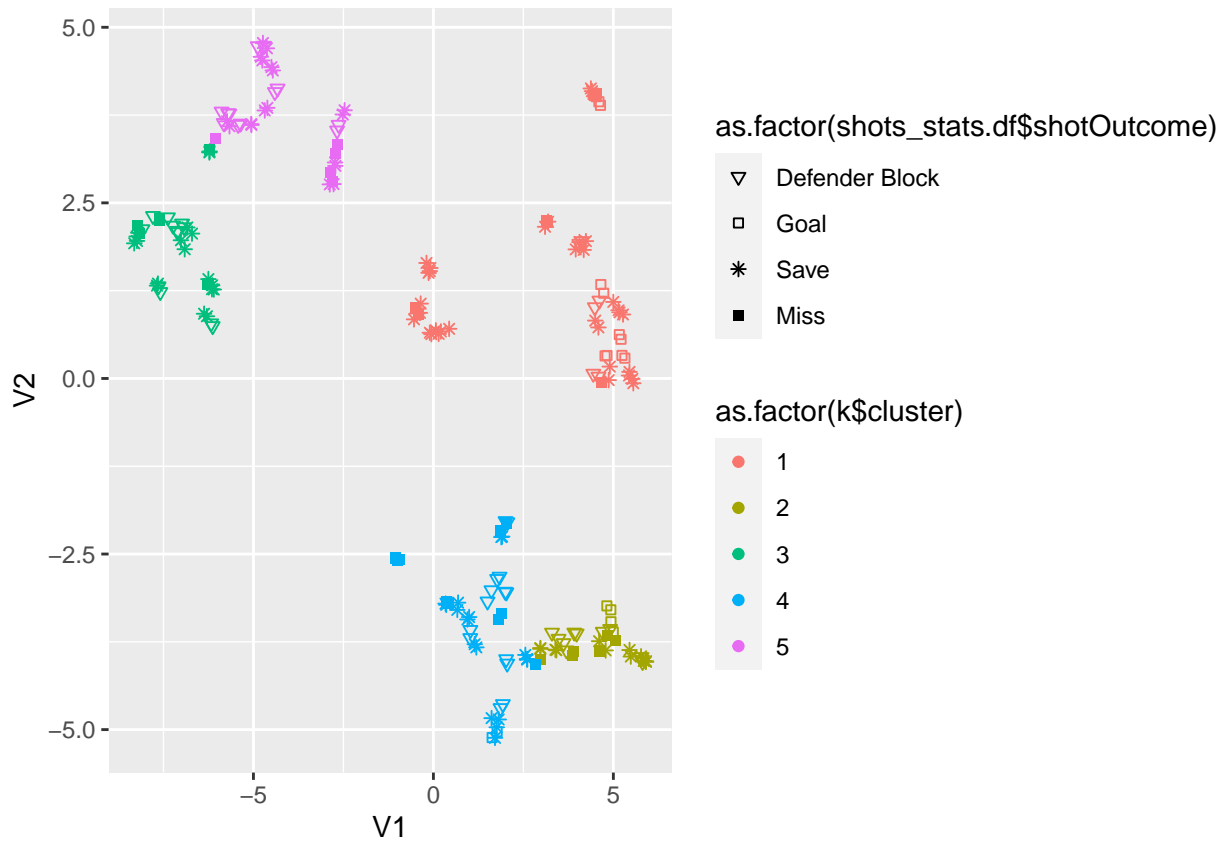
Kind of looks like 5 here, which is consistent with previous results. I'll try the clustering and see if it looks alright, and also generate a summary table to compare the clusters.

```
k <- kmeans(umapCat$layout,5) #4 also looks pretty reasonable
```

```
ggplot()+
```

```
  geom_point(data = as.data.frame(umapCat$layout), aes(x = V1, y = V2, shape = as.factor(shots_stats.df$sh
```

```
  scale_shape_manual(values = c(25,22,8,15,12))
```



*#summary table*

```
tabData <- cbind.data.frame(shots_stats.df[,1:10],k$cluster)
tabData <- cbind.data.frame(tabData,shots_stats.df[,12:13])
head(tabData)
```

```
##   puckDist puckAngle puckSpeed shooterSpeed goalieDist goalieAngle posTime
## 1 688.5900 56.32323 35.60191 12.741100 79.35496 40.29081 2
## 2 370.3595 75.22510 26.85647 10.126463 30.50606 62.68679 15
## 3 379.3229 47.26643 52.76842 18.196121 67.94116 32.18629 1
## 4 408.4321 152.88808 51.62253 13.477262 42.66974 146.41294 39
## 5 278.5743 59.24320 50.14920 20.367026 86.24243 54.11996 60
## 6 606.1261 64.86992 45.44330 5.240912 73.35530 62.38697 37
##   NumOffense NumDefense rightHanded k$cluster defDist defAngle
## 1          1          2          0          3 554.55208 140.29522
## 2          1          2          0          1 77.14952 129.70627
## 3          0          2          0          1 208.92192 125.24775
## 4          0          2          0          2 77.19322 59.64689
## 5          1          2          1          5 129.34843 139.81344
## 6          2          3          0          3 141.06063 169.62383
```

```
tab <- tabData %>% group_by(`k$cluster`) %>% summarise_all(.funs = mean)
kable(tab[,1:9])
```



k\$cluster	puckDist	puckAngle	puckSpeed	shooterSpeed	goalieDist	goalieAngle	posTime	NumOffense
1	207.4834	73.65368	39.20384	15.39178	46.70010	56.00655	19.36667	0.1333333
2	405.4547	116.65869	45.44805	22.05085	82.84534	116.10651	34.84848	0.2424242
3	580.3942	68.41547	37.87735	11.50763	81.53488	61.28986	21.78378	1.1081081
4	479.0101	117.39438	44.46145	13.56194	65.22730	114.58703	24.08889	0.6666667
5	455.3005	54.40157	42.36026	18.35611	86.30226	46.35483	54.74286	0.6571429

```
kable(tab[,10:ncol(tab)])
```

NumDefense	rightHanded	defDist	defAngle
1.333333	0.2333333	141.9952	115.55221
2.212121	0.8787879	112.7941	80.56646
3.108108	0.4594595	194.8992	137.05835
2.022222	0.4666667	288.3022	29.70404
2.085714	0.4857143	198.2589	135.21538

3 is traffic jam, 2 is panic shots, 1 is perfect shots, not sure about 4 and 5. Categorical data seems to focus on angle a lot more. Defender block as a cluster was there because of the defender angle and the shooter angle lining up, so I think the categorizing of all the angles prevented it from forming, as everything became left, center, and right. While defenders being directly in front or 60 degrees to the side are very different for a shooter, both of these scenarios can be placed similarly in the categorized data.

For the UMAP graph: Very interesting, seems the goals all have the same (very high) V1 projection, resulting in them being clustered nicely. I'll take a look at a PCA to see if this is captured linearly as well, as having a more explainable dimensionality reduction would be nice to have

## Discussion of results

*Provide in natural language a clear discussion of your observations.*

3 of the clusters lined up with the clusterings of the continuous, while wide open and defender block were no where to be seen. I explained defender block's absence above, but wide open is probably mixed in with perfect shot for a similar reason to defender block not being present, except this time it is the angle between the goalie and the shooter not being captured instead of the shooter-defender angle.

## Analysis: Question 2: Can relationships between these categorical variables be linearly described?

### Question being asked

*Provide in natural language a statement of what question you're trying to answer*

Can we also see such a clean division with a more explainable dimensionality reduction method?

### Data Preparation

*Provide in natural language a description of the data you are using for this analysis*

*Include a step-by-step description of how you prepare your data for analysis*

*If you're re-using dataframes prepared in another section, simply re-state what data you're using*

Just reusing the dataframe of categorical variables

```
# Include all data processing code (if necessary), clearly commented
```

### Analysis: Methods and Results

*Describe in natural language a statement of the analysis you're trying to do*

Run a PCA and see how it compares with UMAP

Provide clearly commented analysis code; include code for tables and figures!

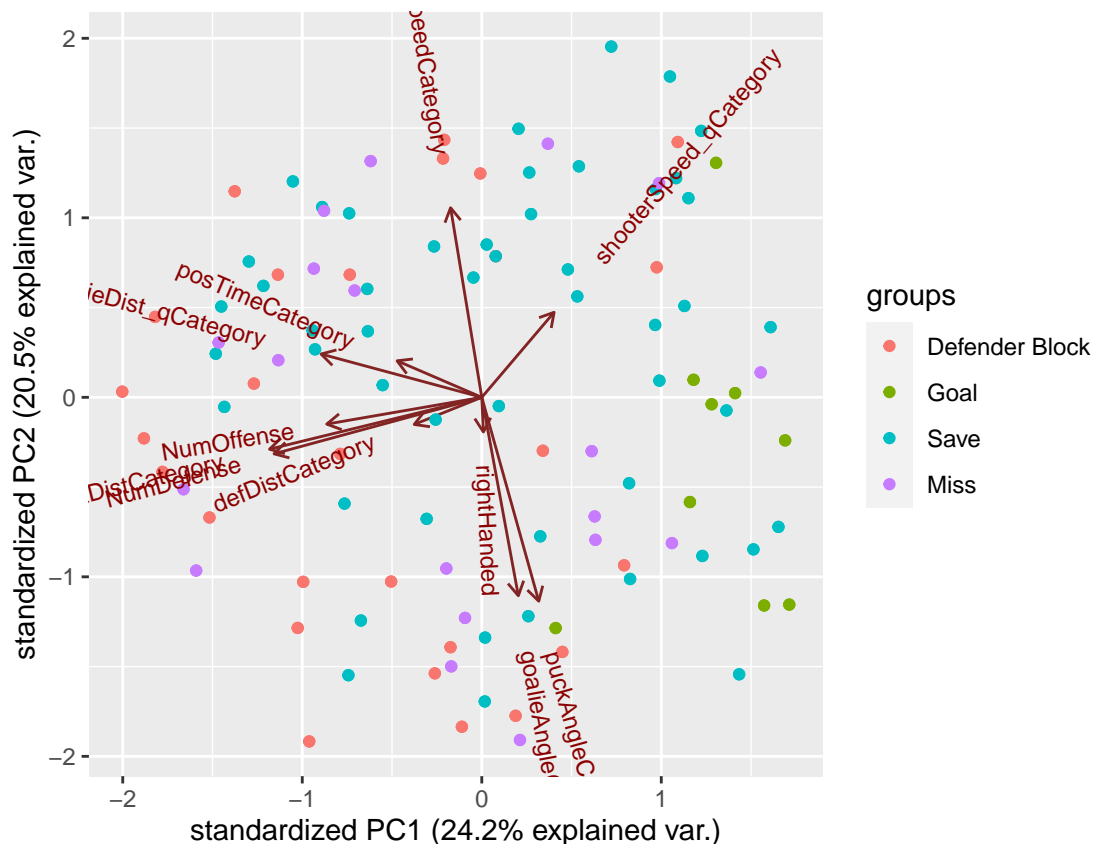
```
# Include all analysis code, clearly commented
# If not possible, screen shots are acceptable.
# If your contributions included things that are not done in an R-notebook,
# (e.g. researching, writing, and coding in Python), you still need to do
# this status notebook in R. Describe what you did here and put any products
# that you created in github. If you are writing online documents (e.g. overleaf
# or google docs), you can include links to the documents in this notebook
# instead of actual text.
```

```
catPCA <- prcomp(justCat)
summary(catPCA)
```

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

```
##          PC1    PC2    PC3    PC4    PC5    PC6    PC7
## Standard deviation  1.2925 1.1886 0.9892 0.8718 0.7126 0.61771 0.5851
## Proportion of Variance 0.2425 0.2051 0.1420 0.1103 0.0737 0.05538 0.0497
## Cumulative Proportion 0.2425 0.4476 0.5896 0.6999 0.7736 0.82898 0.8787
##          PC8    PC9    PC10    PC11
## Standard deviation  0.52525 0.48016 0.43388 0.37569
## Proportion of Variance 0.04005 0.03347 0.02733 0.02049
## Cumulative Proportion 0.91872 0.95219 0.97951 1.00000
```

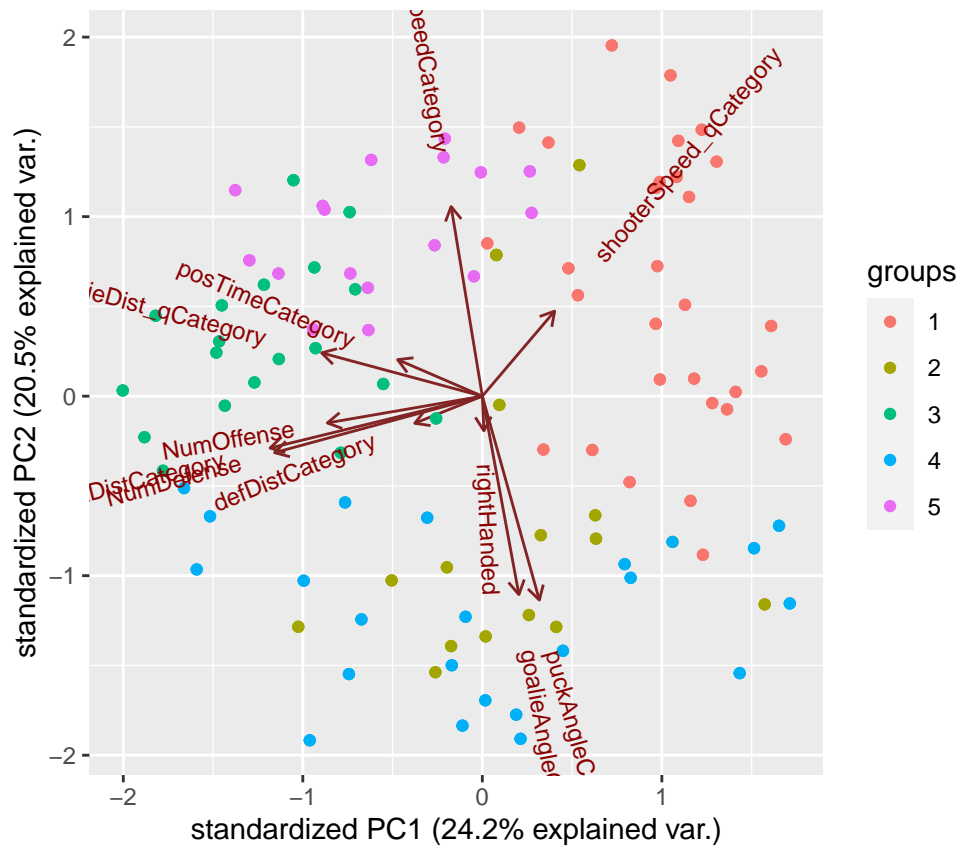
```
ggbiplot(catPCA, groups = as.factor(shots_stats.df$shotOutcome))
```



Goals lining up at a particular value isn't as pronounced as in UMAP, although it is still there. I'll do another

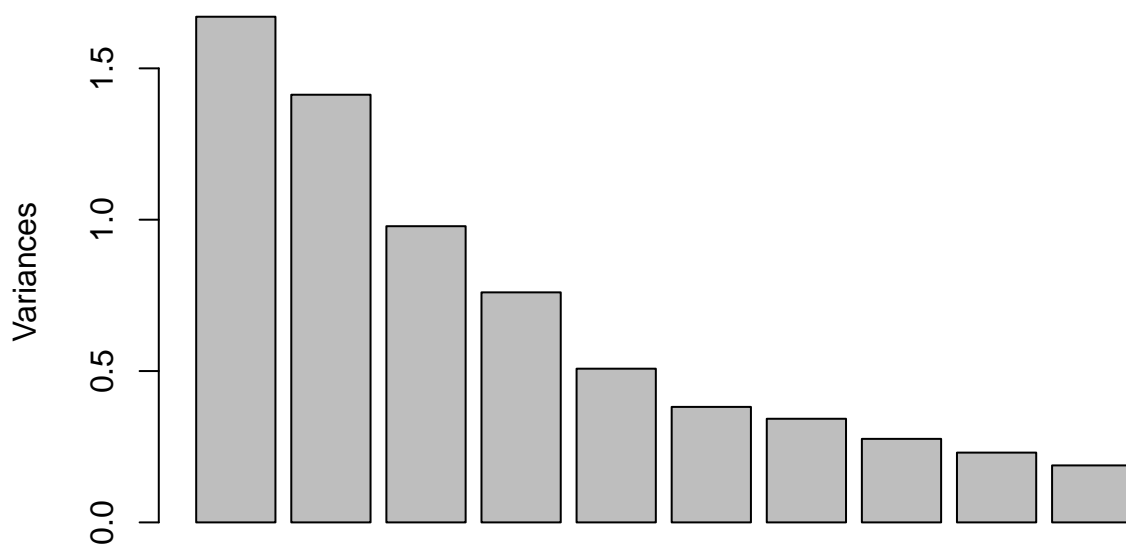
graph to show the UMAP clusters instead of the outcomes, as well as a screeplot

```
ggbiplot(catPCA, groups = as.factor(k$cluster))
```



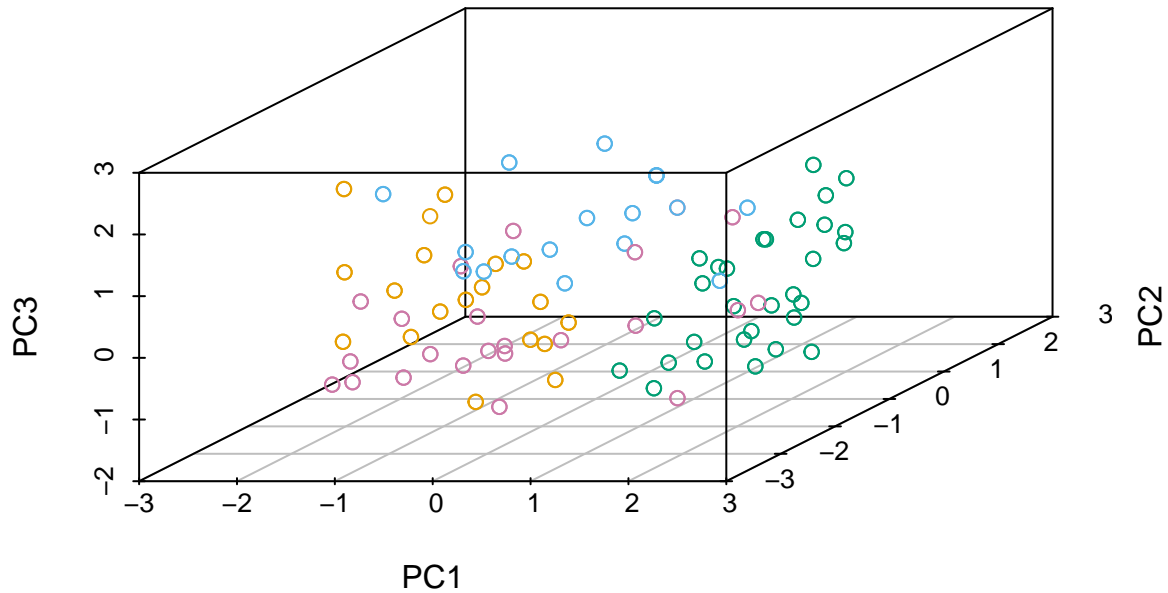
```
screepplot(catPCA)
```

## catPCA



```
colorPalette <- c( rgb(0,0.62,0.45),rgb(86/255,180/255,233/255),  
                  rgb(230/255,159/255,0/255),rgb(204/255,121/255,167/255))
```

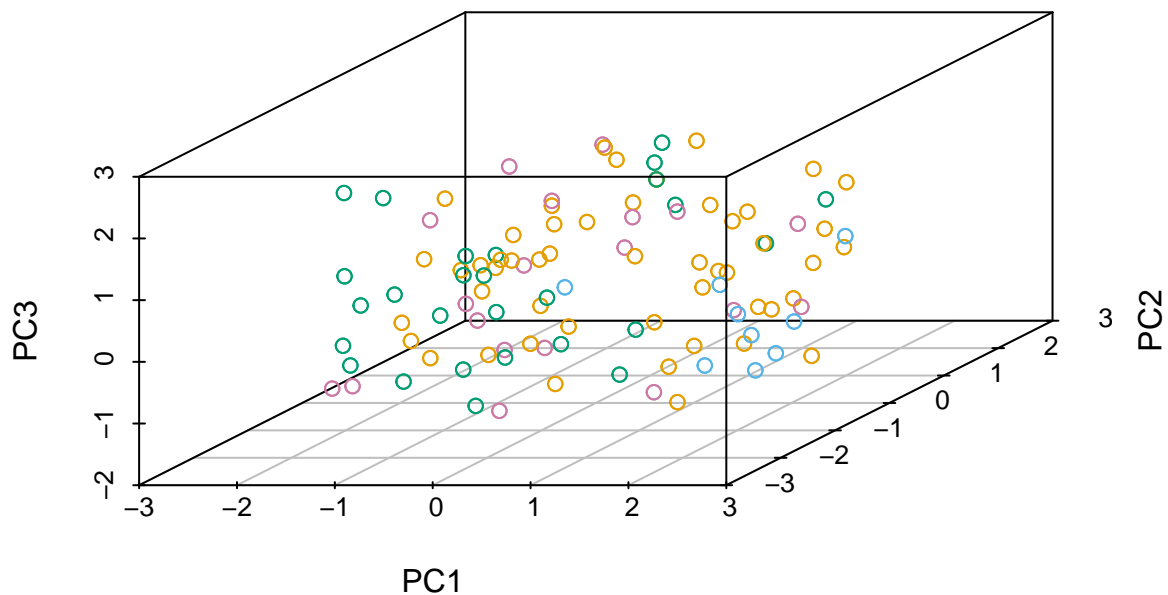
```
colors <- colorPalette[k$cluster]
scatterplot3d(catPCA$x[,1:3],color = colors)
```



```
#Clusters don't nicely align

#s3d <- scatterplot3d(catPCA$x[,1:3],color = colors)
#legend(s3d$xyz.convert(7.5, 3, 4.5), legend = levels(shots_stats.df),
#      col = c("#999999", "#E69F00", "#56B4E9"), pch = 16)

colors2 <- colorPalette[shots_stats.df$outcomes.goal]
scatterplot3d(catPCA$x[,1:3],color = colors2)
```



```
#light blue is goals, yeah doesn't really do much
```

Note the first 3D graph is by cluster, the second is by outcome. While there is still some separation of clusters,

and the UMAP clusters line up alright, it looks like clusters 3 and 5 are getting projected into the same plane in PCA, as are clusters 2 and 4. Unfortunately, going up to 3 dimensions doesn't stop these overlapping clusters. All the features related to players who aren't the shooter or the goalie are heavily correlated with each other and puck distance, which makes sense. Puck angle and goalie angle are also correlated for rather obvious reasons, although the lack of representation right handed is getting in the 2D plot is bizarre. Puck speed is strongly negatively correlated with the angle category, which I wasn't really expecting. Shooter speed is off doing it's own thing.

## Discussion of results

*Provide in natural language a clear discussion of your observations.*

PCA really doesn't do much with the data, there isn't a clear clustering to any great extent. I don't think Jeff's discretization has too much impact on clustering, and I think the fact that all the categoricals got lumped together originally in PCA hasn't changed, as things that were originally categorical are still correlated in this new PCA.

## Analysis: Question 3 (Provide short name)

### Question being asked

*Provide in natural language a statement of what question you're trying to answer*

### Data Preparation

*Provide in natural language a description of the data you are using for this analysis*

*Include a step-by-step description of how you prepare your data for analysis*

*If you're re-using dataframes prepared in another section, simply re-state what data you're using*

*# Include all data processing code (if necessary), clearly commented*

### Analysis methods used

*Describe in natural language a statement of the analysis you're trying to do*

*Provide clearly commented analysis code; include code for tables and figures!*

```
# Include all analysis code, clearly commented
# If not possible, screen shots are acceptable.
# If your contributions included things that are not done in an R-notebook,
# (e.g. researching, writing, and coding in Python), you still need to do
# this status notebook in R. Describe what you did here and put any products
# that you created in github. If you are writing online documents (e.g. overleaf
# or google docs), you can include links to the documents in this notebook
# instead of actual text.
```

## Discussion of results

*Provide in natural language a clear discussion of your observations.*

## Summary and next steps

*Provide in natural language a clear summary and your proposed next steps.*

I'm done with clustering analyses, as the results here for clustering have been dead ends. I'll focus on integrating and finalizing the app, and the final paper for the rest of the semester