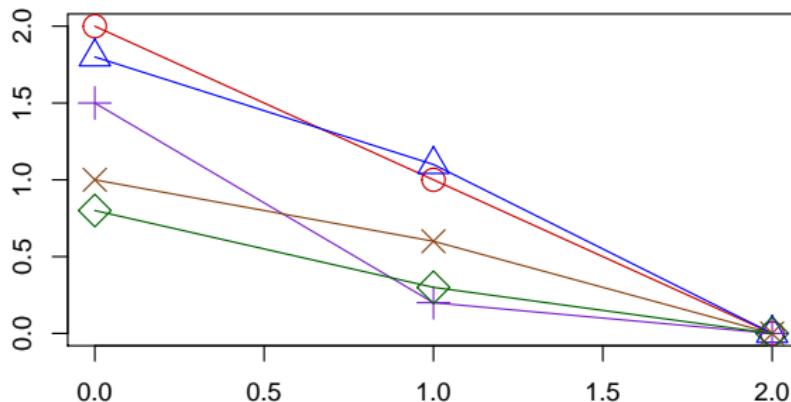


Suppose we have the following feature vectors:

	$f(0)$	$f(1)$	$f(2)$
red	2.0	1.0	0
blue	1.8	1.1	0
purple	1.5	0.2	0
chocolate	1.0	0.6	0
green	0.8	0.3	0

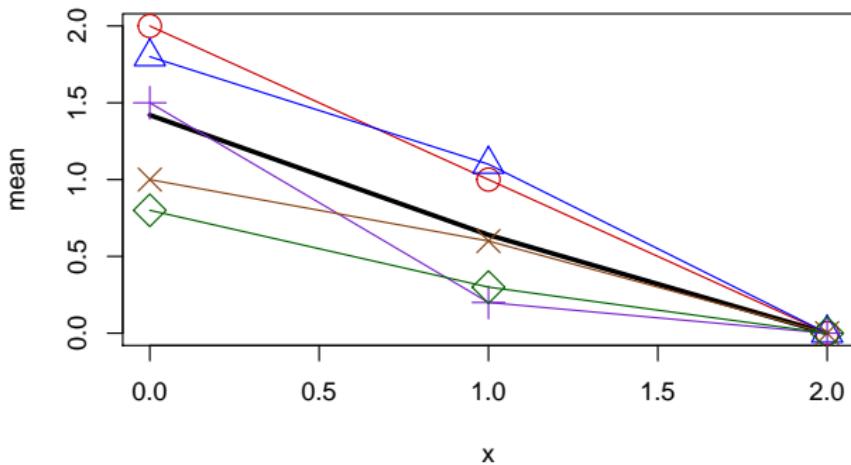
We can plot these feature vectors (regardless of their length) in \mathbb{R}^2



Let's take the mean of our feature vectors and plot it's graph in black:

```
> Y<- rbind(y1, y2, y3, y4, y5)
> mean <- colMeans(Y)
> XY <- cbind(x, mean)
> t(XY)
```

	[,1]	[,2]	[,3]
x	0.00	1.00	2
mean	1.42	0.64	0



To see how these functions (feature vectors) vary about the mean, we will subtract the mean vector from each feature vector:

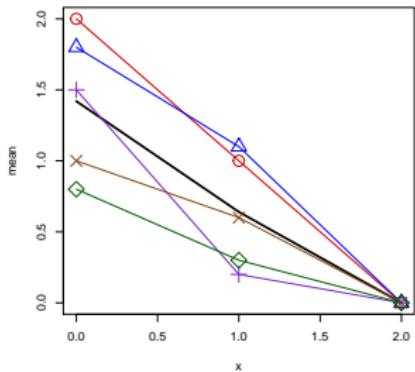


Figure : Raw Data + Mean

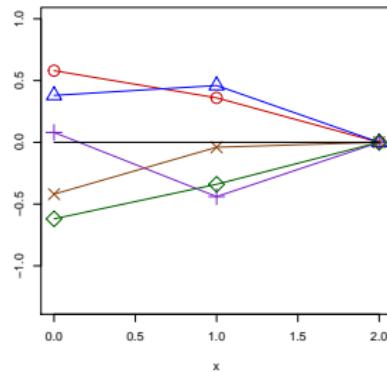
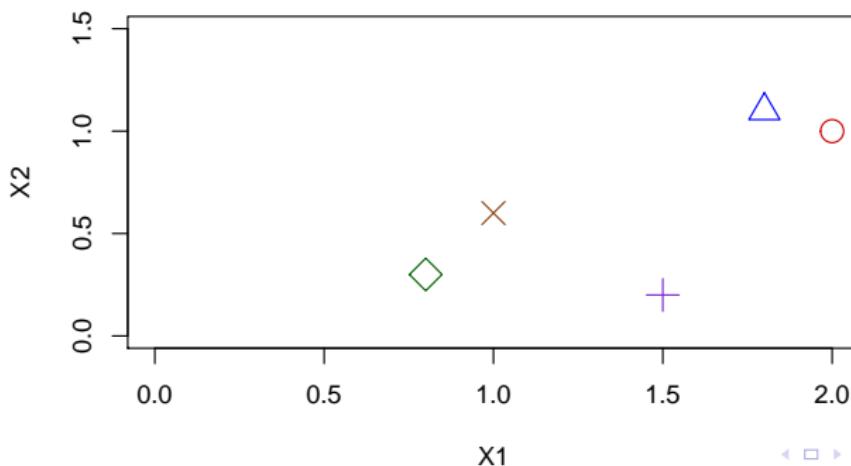


Figure : Mean Residuals

We can also plot our feature vectors in \mathbb{R}^n where n is the number of features

	$f(0)$	$f(1)$	$f(2)$
red	2.0	1.0	0
blue	1.8	1.1	0
purple	1.5	0.2	0
chocolate	1.0	0.6	0
green	0.8	0.3	0



Let's translate these points by subtracting off the mean:

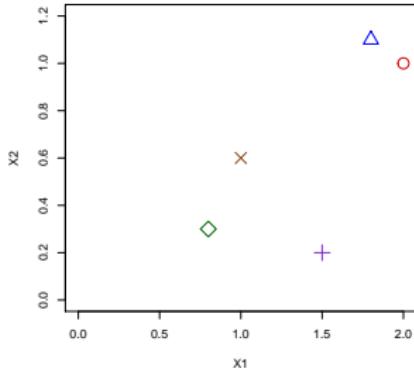


Figure : Raw Data + Mean

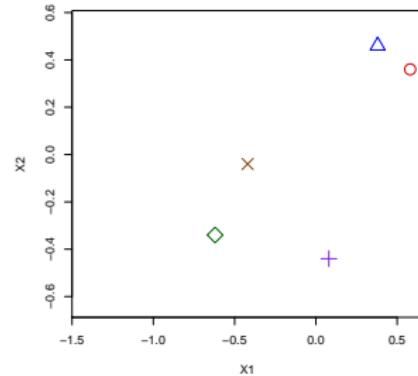


Figure : Centered about the origin

Covariance:

$$Cov(X, Y) = \sum_{i=0}^N \frac{(x_i - \bar{x})(y_i - \bar{y})}{N}$$

	X1	X2
y1	2.0	1.0
y2	1.8	1.1
y3	1.5	0.2
y4	1.0	0.6
y5	0.8	0.3

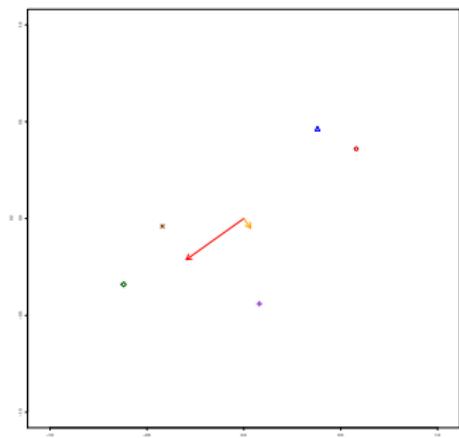
covariance matrix:

	X1	X2
X1	0.262	0.144
X2	0.144	0.163

We can now apply PCA to our data:

Covariance matrix:

	X1	X2
X1	0.262	0.144
X2	0.144	0.163



Eigenvalues and Eigenvectors

\$values

```
[1] 0.36477032 0.06022968
```

\$vectors

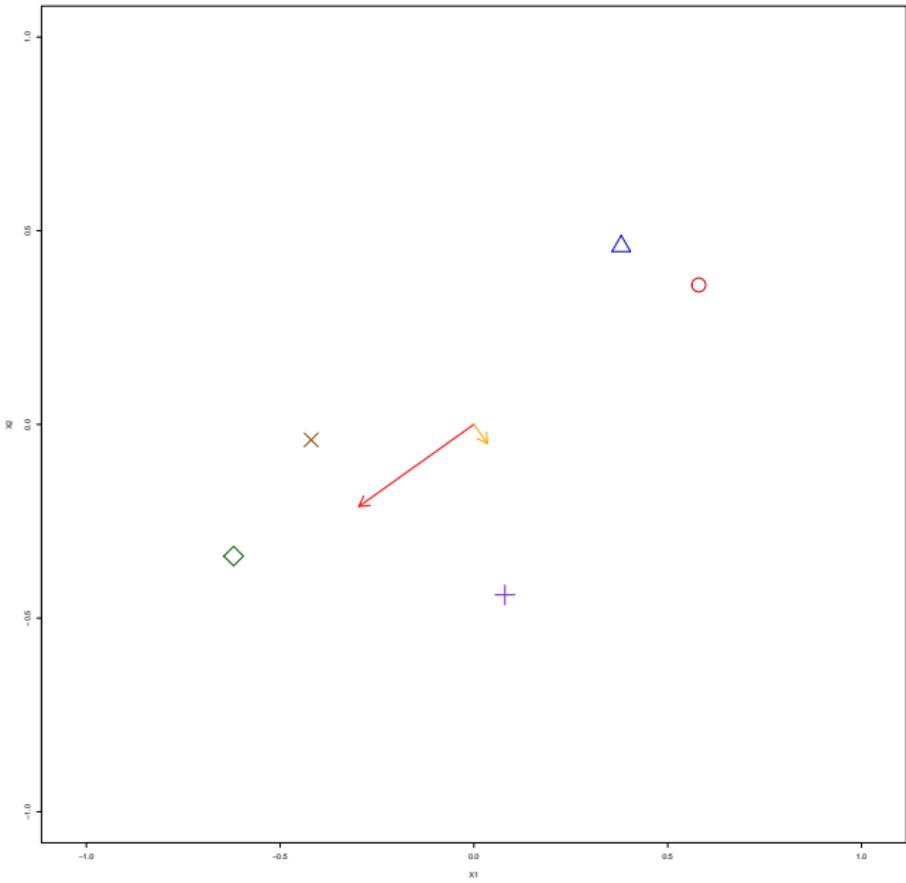
	[,1]	[,2]
[1,]	-0.8139655	0.5809132
[2,]	-0.5809132	-0.8139655

```
> E$values[1]/(E$values[1]+E$values[2])
```

```
[1] 0.8582831
```

```
> sqrt(E$values)
```

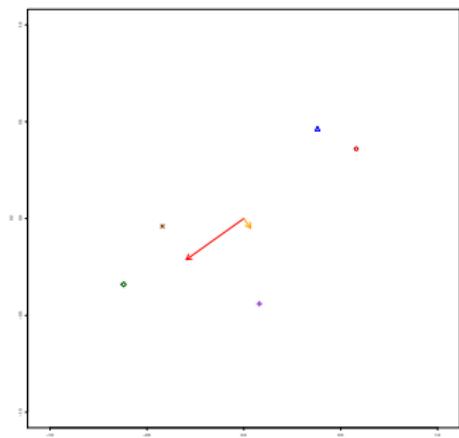
```
[1] 0.6039622 0.2454174
```



We can now apply PCA to our data:

Covariance matrix:

	X1	X2
X1	0.262	0.144
X2	0.144	0.163



Eigenvalues and Eigenvectors

\$values

```
[1] 0.36477032 0.06022968
```

\$vectors

	[,1]	[,2]
[1,]	-0.8139655	0.5809132
[2,]	-0.5809132	-0.8139655

```
> E$values[1]/(E$values[1]+E$values[2])
```

```
[1] 0.8582831
```

```
> sqrt(E$values)
```

```
[1] 0.6039622 0.2454174
```

```
> Y3<- as.matrix(t(t(Y2) - c(mean[1], mean[2])))  
> prin_comp <- prcomp(Y3, scale=FALSE)  
> prin_comp
```

Standard deviations:

```
[1] 0.6039622 0.2454174
```

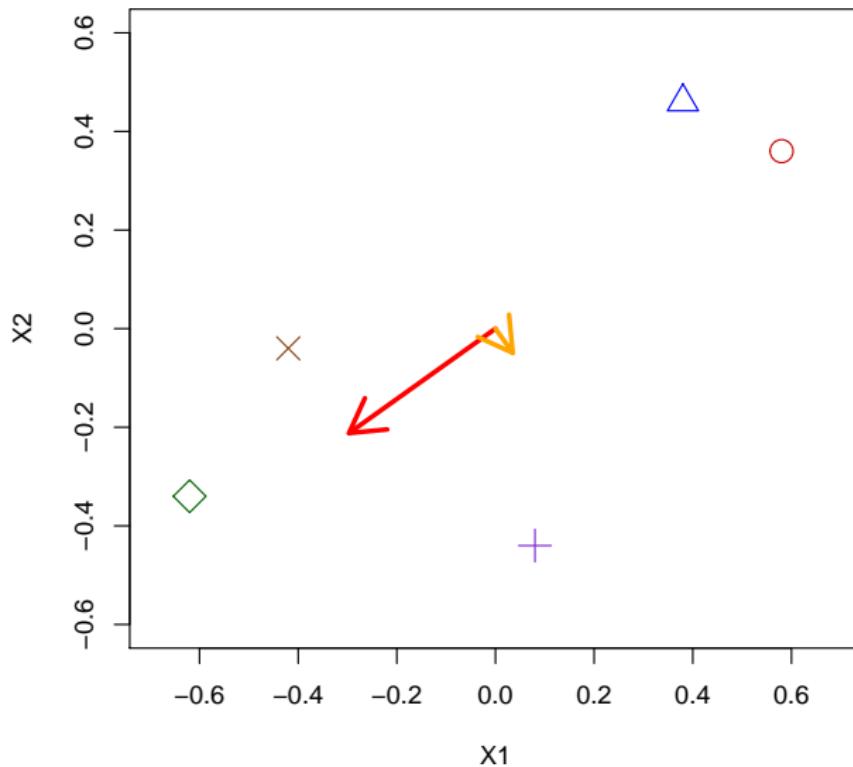
Rotation:

	PC1	PC2
X1	-0.8139655	0.5809132
X2	-0.5809132	-0.8139655

```
> summary(prin_comp)
```

Importance of components:

	PC1	PC2
Standard deviation	0.6040	0.2454
Proportion of Variance	0.8583	0.1417
Cumulative Proportion	0.8583	1.0000



```
[1] 0.6039622 0.2454174
```

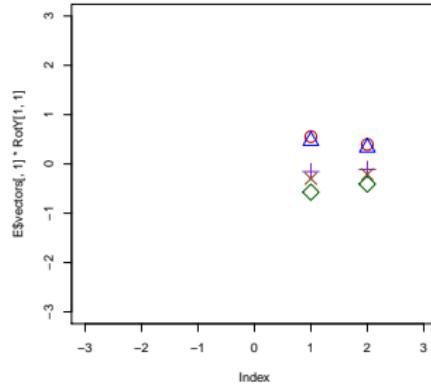
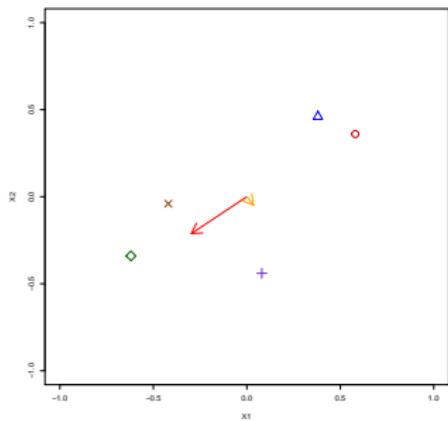
	PC1	PC2
X1	-0.8139655	0.5809132
X2	-0.5809132	-0.8139655

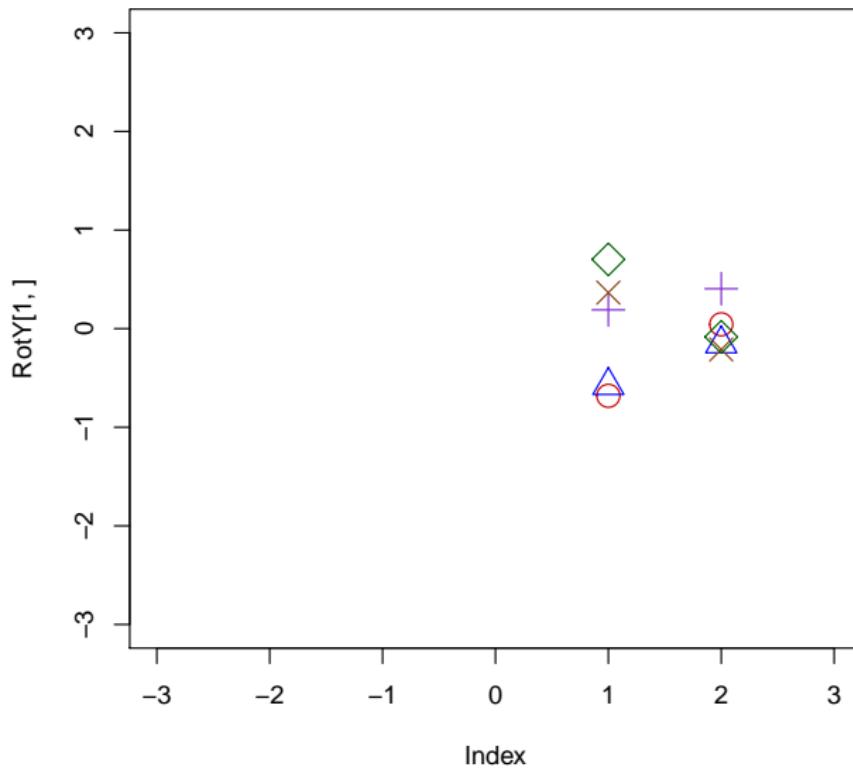
	PC1	PC2
y1	-0.6812288	0.04390205
y2	-0.5765270	-0.15367714
y3	0.1904846	0.40461789
y4	0.3651020	-0.21142491
y5	0.7021691	-0.08341789

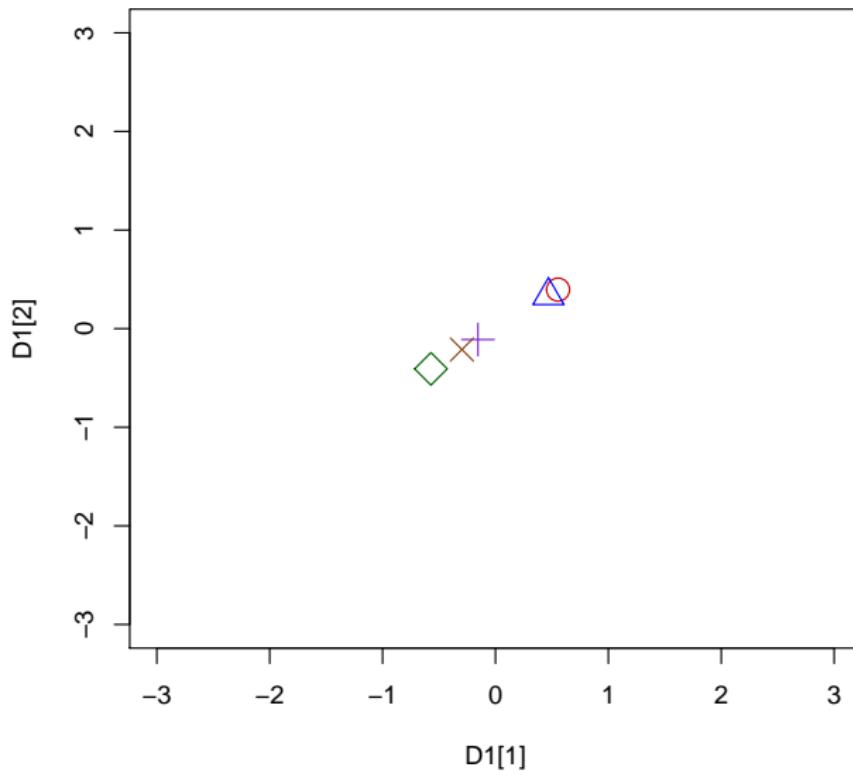
	X1	X2
y1	0.58	0.36
y2	0.38	0.46
y3	0.08	-0.44
y4	-0.42	-0.04
y5	-0.62	-0.34

Covariance matrix:

	X1	X2
X1	0.262	0.144
X2	0.144	0.163



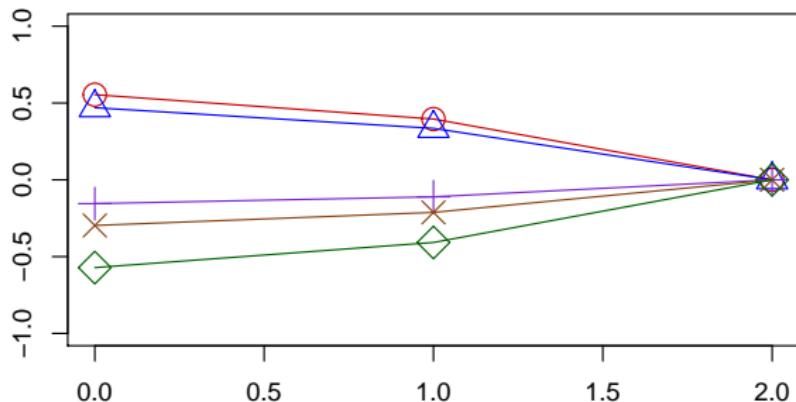




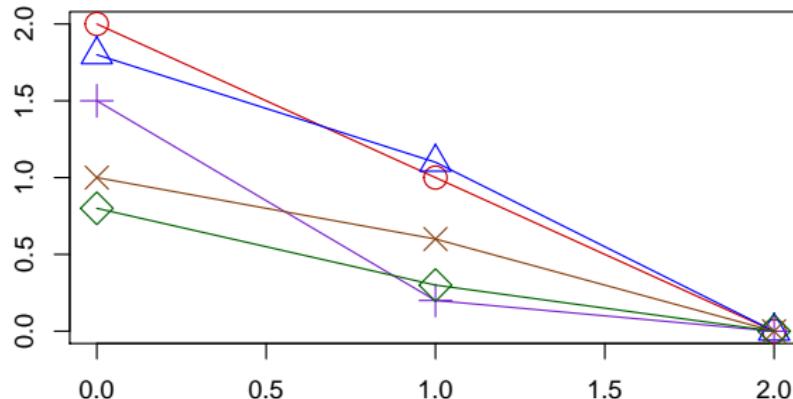
Suppose we have the following feature vectors:

				colorY
[1,]	"0.55449672074308"	"0.395734755466423"	"0"	"red"
[2,]	"0.469273073857219"	"0.334912107110384"	"0"	"blue"
[3,]	"-0.155047861106629"	"-0.110654986955375"	"0"	"purple"
[4,]	"-0.297180483373417"	"-0.212092590483801"	"0"	"chocolate"
[5,]	"-0.571541450120252"	"-0.407899285137631"	"0"	"green"

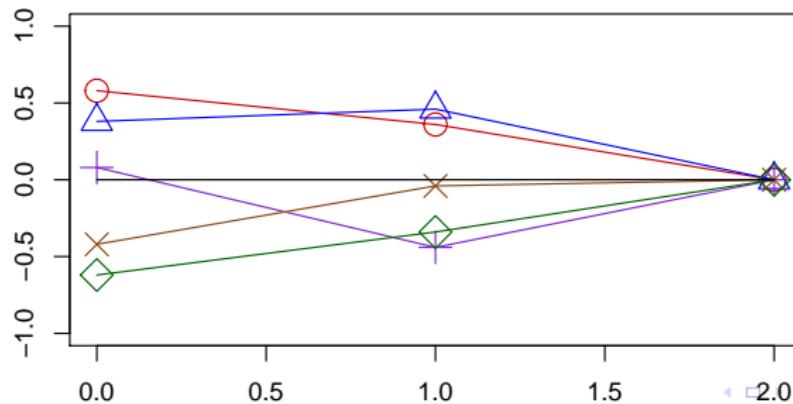
We can plot these feature vectors (regardless of their length) in \mathbb{R}^2



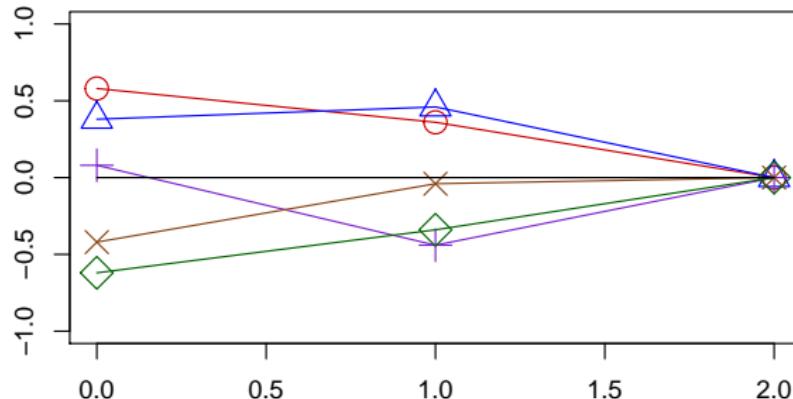
Raw Data



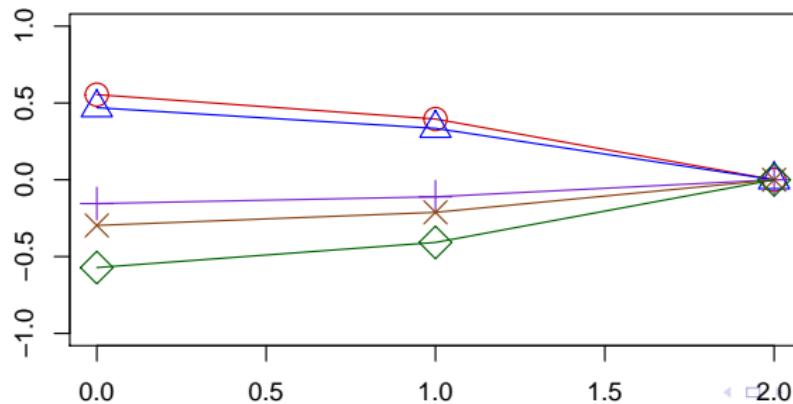
Mean Residual

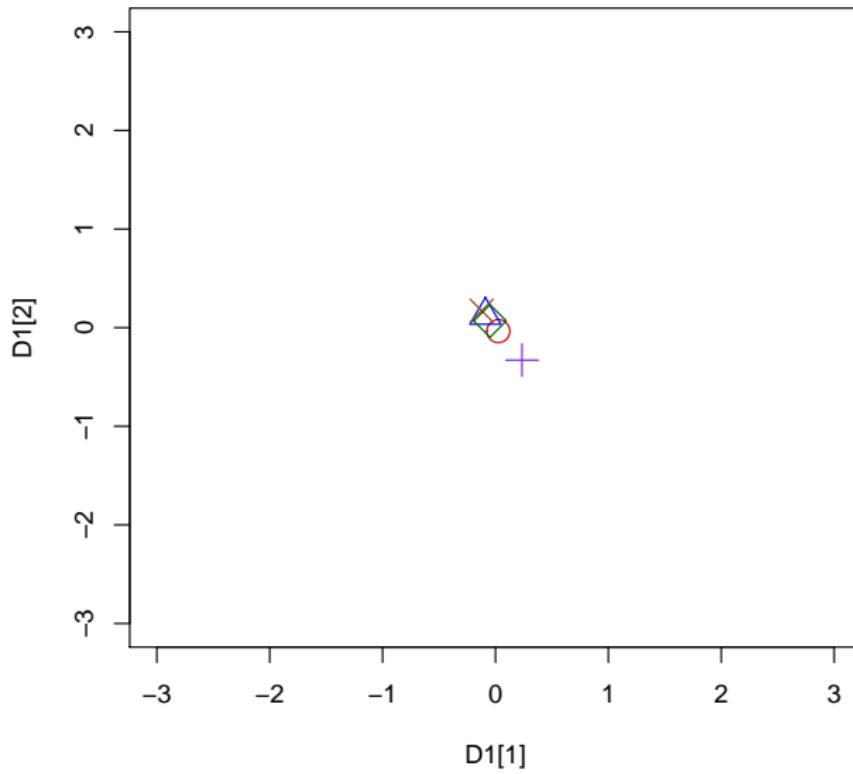


Mean Residual



PC1

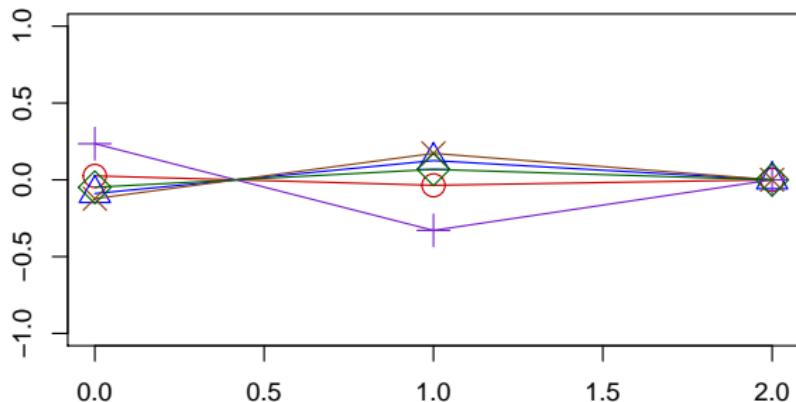




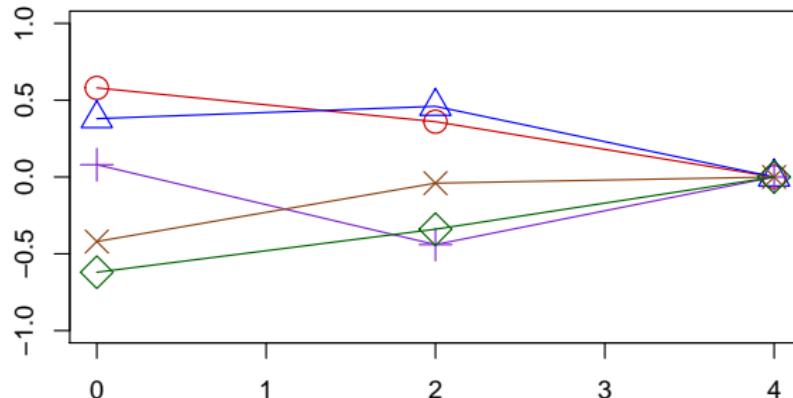
Suppose we have the following feature vectors:

					color
[1,]	"0.0255032792569201"	"-0.0357347554664227"	"0"	"red"	
[2,]	"-0.0892730738572185"	"0.125087892889616"	"0"	"blue"	
[3,]	"0.235047861106629"	"-0.329345013044625"	"0"	"purple"	
[4,]	"-0.122819516626583"	"0.172092590483801"	"0"	"chocolate"	
[5,]	"-0.0484585498797477"	"0.0678992851376303"	"0"	"green"	

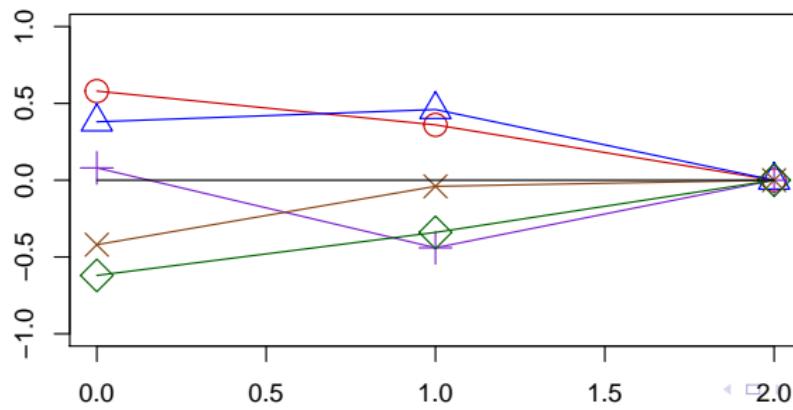
We can plot these feature vectors (regardless of their length) in \mathbb{R}^2

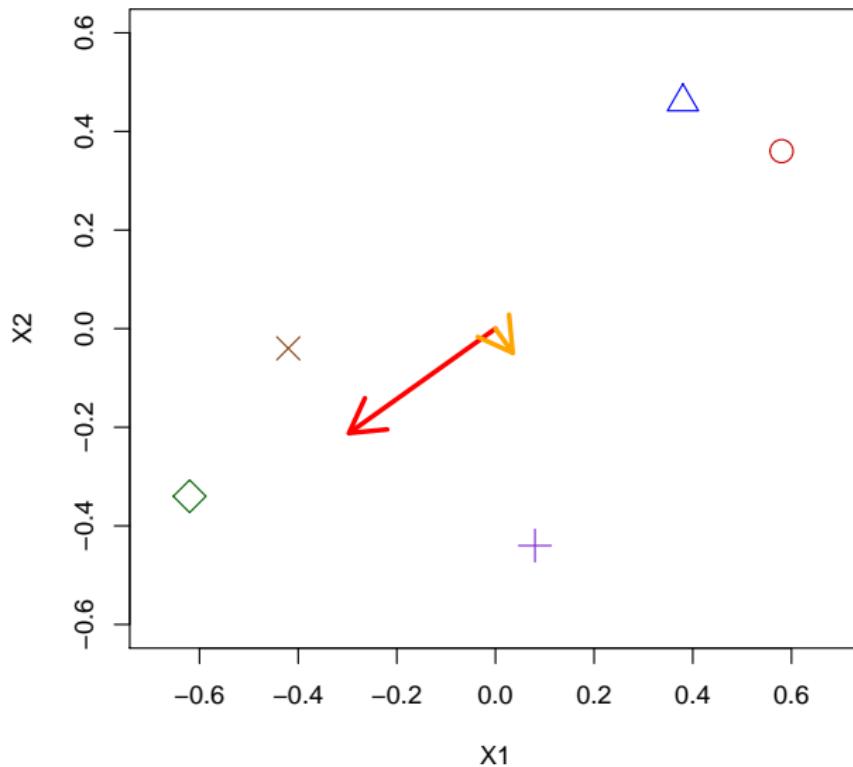


PC1 + PC2

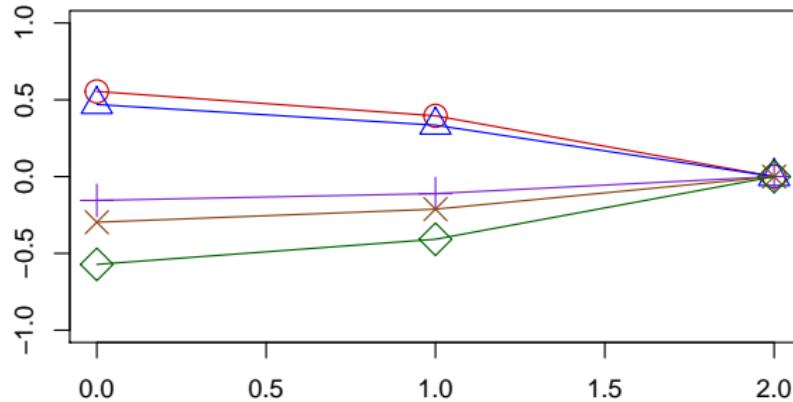


Mean Residual





PC1



PC2

