

diabetesSEM

STAT 360, 4/8/22

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```
diabetesIndicators <- read_excel("diabetesIndicators.xlsx")
```

```
data <- diabetesIndicators %>%  
  mutate(diabetesTF = as.numeric(ifelse(Diabetes_012 > 1, 1, 0)))
```

```
dataT <- data  
dataT$BMI <- 1/dataT$BMI  
dataT$Income <- dataT$Income^4  
dataT$GenHlth <- sqrt(dataT$GenHlth)  
dataT$Education <- dataT$Education^4
```

Because we only have one response variable, we cannot have any clusters, so we don't have any loading matrices/ calculations for that side of things. We remove two variable when we create the explanatory-variable-only matrix because we have two response variables (one generated by the other, we are only going to be using diabetesTF)

```
dat <- dataT[,c(2:22)]  
  
R <- cor(dat)  
eigenVal <- eigen(R)$values  
eigenVec <- eigen(R)$vectors  
d <- sum(eigenVal >= 1)  
d
```

```
## [1] 7
```

Using Kaiser's criteria, we have 7 eigen values above the threshold. In other words, we have 7 factors. We chose Kaiser's criterion because this provided already a substantial number of factors. We did not pick a lower threshold like Joliffe's criterion because, according to Joliffe, there would be 15 factors. This is a massive jump especially since there are only 20 variables in the original dataset (for explanatory variables). To further this point, later in this analysis, adding additional factors was not helpful anyways. As a result, we used Kaiser's criteria.

```
L <- diag(eigenVal[c(1:d)])  
V <- eigenVec[,c(1:d)]  
  
A <- V %*% sqrt(L)
```

```

Z <- scale(dat)
inv <- solve(R)

rotOt <- pca(R, nfactors = d, rotate = "varimax")$loadings[]
rotOb <- pca(R, nfactors = d, rotate = "oblimin")$loadings[]

B0t <- inv %*% rotOt
F0t <- Z %*% B0t

B0b <- inv %*% rotOb
F0b <- Z %*% B0b

cor(F0b)

```

```

##          TC1          TC4          TC6          TC2          TC3          TC7
## TC1  1.000000000  0.212689547 -0.10075782  0.05959143  0.035249031 -0.032686255
## TC4  0.212689547  1.000000000  0.07184993  0.01842221  0.009035975  0.118137617
## TC6 -0.100757817  0.071849933  1.000000000 -0.02019743  0.109414310 -0.171523840
## TC2  0.059591430  0.018422210 -0.02019743  1.000000000 -0.139044115 -0.090739621
## TC3  0.035249031  0.009035975  0.10941431 -0.13904412  1.000000000  0.009541214
## TC7 -0.032686255  0.118137617 -0.17152384 -0.09073962  0.009541214  1.000000000
## TC5  0.009872553 -0.005397148 -0.01468315  0.01175614  0.023834122 -0.060587781
##          TC5
## TC1  0.009872553
## TC4 -0.005397148
## TC6 -0.014683154
## TC2  0.011756140
## TC3  0.023834122
## TC7 -0.060587781
## TC5  1.000000000

```

In class, we decided that if the correlation is above 0.3 or less than -0.3, this means there is a significant correlation between factors. After looking through the correlation matrix above of our loading matrix that has been altered using oblique rotation, there are no correlations that are less than -0.3 or greater than 0.3. Although there are correlations that are close to this threshold, none of them pass, so an oblique rotation is not needed.

```

d2 <- d + 2
rotOt1 <- pca(R, nfactors = d2, rotate = "varimax")$loadings[]
print("Orthogonally rotated loading matrix using 7 factors (as per Kaiser's criterion)")

```

```
## [1] "Orthogonally rotated loading matrix using 7 factors (as per Kaiser's criterion)"
```

```
rotOt
```

```

##          RC1          RC6          RC4          RC3
## HighBP      0.08852025  0.686045289 -0.174682097  0.009717405
## HighChol    0.05959685  0.588903040 -0.059437429  0.056203091
## CholCheck   0.08963070  0.256238847 -0.004862112  0.072398869
## BMI        -0.18809658 -0.629837271 -0.103994986  0.211821118
## Smoker      0.06699898  0.110510187 -0.150953589 -0.094850892
## Stroke      0.17518058 -0.056644497 -0.024787186 -0.023646107

```

## HeartDiseaseorAttack	0.13843656	0.230649896	-0.004722441	-0.009345548
## PhysActivity	-0.31740655	-0.089597764	0.294237509	0.361988128
## Fruits	-0.05966460	-0.012171188	-0.016584071	0.720539050
## Veggies	-0.02119845	0.041636368	0.173613387	0.701681742
## HvyAlcoholConsump	0.04029255	-0.048758162	0.093286973	0.033022560
## AnyHealthcare	0.06221350	-0.004470572	0.168419302	-0.028597741
## NoDocbcCost	0.29320118	0.028509248	-0.086046032	0.093508465
## GenHlth	0.59693544	0.347182582	-0.215989214	-0.133955151
## MentHlth	0.69162626	-0.067889832	0.035107507	-0.024923564
## PhysHlth	0.76437092	0.075374833	-0.070806261	-0.022799960
## DiffWalk	0.62677443	0.186803379	-0.217369434	-0.037452079
## Sex	-0.21470183	0.283107600	0.464397706	-0.342683883
## Age	-0.09603335	0.368460377	-0.371734013	0.157641575
## Education	-0.05778701	-0.132403265	0.673588885	0.239583564
## Income	-0.21368805	-0.098200011	0.721285442	0.086626957
##	RC2	RC7	RC5	
## HighBP	0.08978591	0.15785970	0.052385599	
## HighChol	0.09857905	0.14991066	0.105489020	
## CholCheck	0.39285125	-0.15513519	-0.067195130	
## BMI	0.14773303	0.24734074	0.200694301	
## Smoker	-0.08130963	0.18542771	0.648190894	
## Stroke	-0.00474838	0.65959069	-0.062519266	
## HeartDiseaseorAttack	-0.02530557	0.64309613	0.038294117	
## PhysActivity	-0.08634954	0.02074129	0.011397249	
## Fruits	0.01817828	0.04821077	-0.118044001	
## Veggies	-0.04380348	-0.06970281	0.076926418	
## HvyAlcoholConsump	0.03831142	-0.18935796	0.767478195	
## AnyHealthcare	0.72965494	0.02912816	0.004509130	
## NoDocbcCost	-0.60542107	-0.05501506	0.003546362	
## GenHlth	-0.07449888	0.17466919	0.015831049	
## MentHlth	-0.13620648	-0.06477537	0.112377186	
## PhysHlth	0.01472056	0.18922633	0.017685028	
## DiffWalk	0.07580036	0.22092070	-0.031884824	
## Sex	-0.23367106	0.26895963	0.130851592	
## Age	0.38086617	0.39186222	0.103292932	
## Education	0.18342794	-0.01223731	-0.095031039	
## Income	0.21239849	-0.10274809	0.006436500	

```
print("Orthogonally rotated loading matrix using 9 factors")
```

```
## [1] "Orthogonally rotated loading matrix using 9 factors"
```

```
rotOt1
```

##	RC1	RC8	RC4	RC2
## HighBP	0.14270945	0.73289946	-0.12233981	0.016871740
## HighChol	0.06509643	0.70677749	0.05400438	-0.052208125
## CholCheck	0.01910859	0.11105054	-0.02954135	0.114339485
## BMI	-0.27208773	-0.41518781	-0.01340410	0.101631354
## Smoker	0.14515570	-0.02371331	-0.43843703	0.087807771
## Stroke	0.15651951	0.07919454	0.02479704	-0.057579691
## HeartDiseaseorAttack	0.18436155	0.26046116	-0.07169564	0.005211554
## PhysActivity	-0.37756479	-0.11592857	0.25079741	-0.150446984

## Fruits	-0.05356347	-0.01108549	0.01345132	0.038975770
## Veggies	-0.01100364	-0.01168856	0.13801486	-0.004289220
## HvyAlcoholConsump	-0.01632793	0.09105835	0.12183398	-0.051439152
## AnyHealthcare	0.10718167	0.02150934	0.20001002	0.712010415
## NoDocbcCost	0.20311135	0.01218712	-0.04701914	-0.727415375
## GenHlth	0.64733525	0.27468956	-0.24476955	-0.078313028
## MentHlth	0.64137596	-0.11276392	0.06241112	-0.239072753
## PhysHlth	0.78773157	0.03094379	-0.06505575	-0.001214180
## DiffWalk	0.67529897	0.18354119	-0.18205665	0.078197020
## Sex	-0.10331733	0.01909606	0.09861330	0.003667330
## Age	-0.03140443	0.55083605	-0.27349561	0.362128852
## Education	-0.11060916	-0.04764983	0.77740565	0.091852574
## Income	-0.22854922	-0.11419016	0.69935942	0.205502803
##	RC3	RC7	RC6	RC5
## HighBP	-0.017519281	0.042117787	0.071926044	-0.013928224
## HighChol	-0.053060207	0.108978872	-0.020821632	0.095210351
## CholCheck	-0.006286401	-0.004757031	-0.028608631	-0.043879322
## BMI	0.098879481	0.437536005	-0.335912229	0.283380769
## Smoker	0.088689309	0.093800129	0.425158948	0.505729305
## Stroke	-0.093230603	0.703540178	-0.012752391	-0.067458810
## HeartDiseaseorAttack	0.025511590	0.574532135	0.237780765	-0.044943989
## PhysActivity	0.319854740	0.116772948	0.074790253	0.045974438
## Fruits	0.750002462	0.032796248	-0.128590829	-0.119400233
## Veggies	0.760490252	-0.104974701	0.022218567	0.073751142
## HvyAlcoholConsump	-0.068992291	-0.100788443	-0.069842903	0.831365165
## AnyHealthcare	-0.002948904	-0.001450823	-0.053730492	0.020718585
## NoDocbcCost	-0.004870303	0.044092011	-0.066217220	0.042817524
## GenHlth	-0.077236781	0.077802847	0.099104629	-0.053545809
## MentHlth	-0.062630705	-0.001026385	-0.077481881	0.162104839
## PhysHlth	0.017477105	0.143853587	-0.005649783	0.001439319
## DiffWalk	0.002932505	0.137496513	-0.046151642	-0.071803381
## Sex	-0.104934153	0.080773882	0.833361007	-0.025753374
## Age	0.128686576	0.309238120	-0.132292310	0.039823333
## Education	0.141819521	0.044971100	-0.031904391	0.024098208
## Income	0.077712155	-0.093699591	0.176973116	0.076610608
##	RC9			
## HighBP	0.038049353			
## HighChol	0.072984620			
## CholCheck	0.938136904			
## BMI	-0.004293975			
## Smoker	0.036714719			
## Stroke	0.005712239			
## HeartDiseaseorAttack	0.002805088			
## PhysActivity	0.235320610			
## Fruits	-0.002442106			
## Veggies	-0.014597471			
## HvyAlcoholConsump	-0.048936948			
## AnyHealthcare	0.173318380			
## NoDocbcCost	0.041761089			
## GenHlth	0.022407109			
## MentHlth	0.107408990			
## PhysHlth	0.003395973			
## DiffWalk	-0.060864460			
## Sex	-0.029506297			

```
## Age -0.047601474
## Education -0.026705790
## Income 0.014189784
```

After performing an orthogonal rotation to get our optimal rotated loading matrix, we found that we had several variables that were complex. We partially resolved this by adding 2 additional factors. After this, we still had four variables that load onto more than one factor. These four variables are BMI, Smoker, PhysActivity, and Age. We were unable to find an amount of factors (within reason) that could contain more variables without complexity, therefore, we decided to retain the mentioned complex variables as individual measured variables for further analysis.

```
rotOtTrim <- rotOt1[c(1,2,3,6,7,8,9,10,11,12,13,14,15,16,17,18,20,21),]
rotOtTrim2 <- rotOtTrim^2
var <- 0
vec <- c()
for(i in 1:d2)
{
  vec <- append(vec, ((sum(rotOtTrim2[,i]) + (4/d2)) / 21))
  var <- var + ((sum(rotOtTrim2[,i]) + (4/d2)) / 21)
}
varOt <- pca(R, nfactors = d2, rotate = "varimax")$Vaccounted
varOt[2,]
```

```
##          RC1          RC8          RC4          RC2          RC3          RC7          RC6
## 0.11239037 0.08383690 0.07787031 0.06436187 0.06390579 0.05882900 0.05494351
##          RC5          RC9
## 0.05295542 0.04751719
```

```
vec
```

```
## [1] 0.12897877 0.08231694 0.08631021 0.07842225 0.08344109 0.06590424 0.06129332
## [8] 0.05804074 0.06850825
```

```
var
```

```
## [1] 0.7132158
```

Because we removed some of our variables from our set of factors, we manually calculated the variance retained. Above you can see the original variance, variance with the 4 removed variables accounted for, and the sum of variance explained with our manual calculation. With our 9 factors and 4 individual variables, we account for ~71.32% of the variance within our original variables.

```
commOt <- pca(R, nfactors = d2, rotate = "varimax")$communality
commOt
```

```
##          HighBP          HighChol          CholCheck
##          0.5816552          0.5389312          0.9095504
##          BMI          Smoker          Stroke
##          0.6512963          0.6761042          0.5431071
## HeartDiseaseorAttack          PhysActivity          Fruits
```

##	0.4963022	0.4205542	0.5990692
##	Veggies	HvyAlcoholConsump	AnyHealthcare
##	0.6148353	0.7394069	0.5922797
##	NoDocbcCost	GenHlth	MentHlth
##	0.5826765	0.5857521	0.5328715
##	PhysHlth	DiffWalk	Sex
##	0.6467571	0.5588795	0.7343372
##	Age	Education	Income
##	0.6438852	0.6517483	0.6488179

58.17% of the variance in HighBP is retained with our dimensionality reduction 53.89% of the variance in HighChol is retained with our dimensionality reduction 90.96% of the variance in CholCheck is retained with our dimensionality reduction 65.13% of the variance in BMI is retained with our dimensionality reduction 67.61% of the variance in Smoker is retained with our dimensionality reduction 54.31% of the variance in Stroke is retained with our dimensionality reduction 49.63% of the variance in HeartDiseaseorAttack is retained with our dimensionality reduction 42.06% of the variance in PhysActivity is retained with our dimensionality reduction 59.91% of the variance in Fruits is retained with our dimensionality reduction 61.48% of the variance in Veggies is retained with our dimensionality reduction 73.94% of the variance in HvyAlcoholConsump is retained with our dimensionality reduction 59.23% of the variance in AnyHealthCare is retained with our dimensionality reduction

58.27% of the variance in NoDocbsCost is retained with our dimensionality reduction

58.58% of the variance in GenHlth is retained with our dimensionality reduction

53.29% of the variance in MentHlth is retained with our dimensionality reduction

64.68% of the variance in PhysHlth is retained with our dimensionality reduction

55.89% of the variance in DiffWalk is retained with our dimensionality reduction 73.43% of the variance in Sex is retained with our dimensionality reduction 64.39% of the variance in Age is retained with our dimensionality reduction

65.17% of the variance in Education is retained with our dimensionality reduction

64.88% of the variance of Income is retained with our dimensionality reduction

```
#EQN1 <- '
#      HEALTH =~ GenHlth + PhysHlth + MentHlth + DiffWalk
#      HEALTHCARE =~ NoDocbcCost + AnyHealthcare
#      DIET =~ Fruits + Veggies
#      CLASS =~ Education + Income
#      SUBSTANCE =~ HvyAlcoholConsump
#      BODY =~ Sex
#      HISTORY =~ Stroke + HeartDiseaseorAttack
#      BLOOD =~ HighBP + HighChol
#      diabetesTF ~ HEALTH + HEALTHCARE + DIET + CLASS + SUBSTANCE + BODY + HISTORY + BLOOD +
#      '

EQN1 <- '
HEALTH =~ GenHlth + PhysHlth + MentHlth + DiffWalk
HEALTHCARE =~ NoDocbcCost + AnyHealthcare
DIET =~ Fruits + Veggies
CLASS =~ Education + Income
HISTORY =~ Stroke + HeartDiseaseorAttack
BLOOD =~ HighBP + HighChol
diabetesTF ~ HEALTH + HEALTHCARE + DIET + CLASS + HISTORY + BLOOD + CholCheck + BMI + Smoker
'
```

```
library(lavaan)
```

```
## Warning: package 'lavaan' was built under R version 4.1.2
```

```
## This is lavaan 0.6-11
```

```
## lavaan is FREE software! Please report any bugs.
```

```
##
```

```
## Attaching package: 'lavaan'
```

```
## The following object is masked from 'package:psych':
```

```
##
```

```
## cor2cov
```

```
M1 <- sem(model = EQN1, data = scale(dataT))
parameterEstimates(M1)
```

##	lhs	op	rhs	est	se	z	pvalue
## 1	HEALTH	=~	GenHlth	1.000	0.000	NA	NA
## 2	HEALTH	=~	PhysHlth	0.930	0.003	269.966	0.000
## 3	HEALTH	=~	MentHlth	0.564	0.003	176.758	0.000
## 4	HEALTH	=~	DiffWalk	0.876	0.003	259.482	0.000
## 5	HEALTHCARE	=~	NoDocbcCost	1.000	0.000	NA	NA
## 6	HEALTHCARE	=~	AnyHealthcare	-0.636	0.009	-70.302	0.000
## 7	DIET	=~	Fruits	1.000	0.000	NA	NA
## 8	DIET	=~	Veggies	1.551	0.028	55.202	0.000
## 9	CLASS	=~	Education	1.000	0.000	NA	NA
## 10	CLASS	=~	Income	1.369	0.008	163.722	0.000
## 11	HISTORY	=~	Stroke	1.000	0.000	NA	NA
## 12	HISTORY	=~	HeartDiseaseorAttack	1.443	0.014	102.057	0.000
## 13	BLOOD	=~	HighBP	1.000	0.000	NA	NA
## 14	BLOOD	=~	HighChol	0.737	0.006	126.242	0.000
## 15	diabetesTF	~	HEALTH	0.157	0.006	27.628	0.000
## 16	diabetesTF	~	HEALTHCARE	-0.007	0.006	-1.254	0.210
## 17	diabetesTF	~	DIET	0.004	0.007	0.533	0.594
## 18	diabetesTF	~	CLASS	-0.033	0.006	-5.314	0.000
## 19	diabetesTF	~	HISTORY	0.160	0.015	10.606	0.000
## 20	diabetesTF	~	BLOOD	0.355	0.008	44.149	0.000
## 21	diabetesTF	~	CholCheck	0.025	0.002	13.434	0.000
## 22	diabetesTF	~	BMI	-0.138	0.002	-74.181	0.000
## 23	diabetesTF	~	Smoker	-0.008	0.002	-4.354	0.000
## 24	diabetesTF	~	PhysActivity	-0.010	0.002	-5.643	0.000
## 25	diabetesTF	~	Age	0.067	0.002	36.207	0.000
## 26	diabetesTF	~	Sex	0.021	0.002	11.367	0.000
## 27	diabetesTF	~	HvyAlcoholConsump	-0.035	0.002	-19.173	0.000
## 28	GenHlth	~~	GenHlth	0.470	0.002	233.308	0.000
## 29	PhysHlth	~~	PhysHlth	0.541	0.002	263.556	0.000
## 30	MentHlth	~~	MentHlth	0.831	0.002	334.824	0.000
## 31	DiffWalk	~~	DiffWalk	0.593	0.002	281.853	0.000
## 32	NoDocbcCost	~~	NoDocbcCost	0.634	0.006	112.938	0.000
## 33	AnyHealthcare	~~	AnyHealthcare	0.852	0.003	264.693	0.000

## 34	Fruits	~~	Fruits	0.836	0.004	219.880	0.000
## 35	Veggies	~~	Veggies	0.605	0.007	81.851	0.000
## 36	Education	~~	Education	0.675	0.003	255.946	0.000
## 37	Income	~~	Income	0.391	0.004	108.478	0.000
## 38	Stroke	~~	Stroke	0.859	0.003	299.947	0.000
## 39	HeartDiseaseorAttack	~~	HeartDiseaseorAttack	0.707	0.004	187.092	0.000
## 40	HighBP	~~	HighBP	0.595	0.004	163.829	0.000
## 41	HighChol	~~	HighChol	0.780	0.003	278.192	0.000
## 42	diabetesTF	~~	diabetesTF	0.809	0.002	328.640	0.000
## 43	HEALTH	~~	HEALTH	0.530	0.003	182.300	0.000
## 44	HEALTHCARE	~~	HEALTHCARE	0.366	0.006	63.574	0.000
## 45	DIET	~~	DIET	0.164	0.003	48.729	0.000
## 46	CLASS	~~	CLASS	0.325	0.003	117.457	0.000
## 47	HISTORY	~~	HISTORY	0.141	0.002	66.707	0.000
## 48	BLOOD	~~	BLOOD	0.405	0.004	102.807	0.000
## 49	HEALTH	~~	HEALTHCARE	0.136	0.002	82.103	0.000
## 50	HEALTH	~~	DIET	-0.066	0.001	-48.194	0.000
## 51	HEALTH	~~	CLASS	-0.224	0.002	-132.921	0.000
## 52	HEALTH	~~	HISTORY	0.158	0.002	99.822	0.000
## 53	HEALTH	~~	BLOOD	0.236	0.002	135.070	0.000
## 54	HEALTHCARE	~~	DIET	-0.028	0.001	-24.059	0.000
## 55	HEALTHCARE	~~	CLASS	-0.139	0.002	-90.866	0.000
## 56	HEALTHCARE	~~	HISTORY	0.015	0.001	13.231	0.000
## 57	HEALTHCARE	~~	BLOOD	-0.002	0.002	-1.495	0.135
## 58	DIET	~~	CLASS	0.073	0.001	53.727	0.000
## 59	DIET	~~	HISTORY	-0.019	0.001	-23.458	0.000
## 60	DIET	~~	BLOOD	-0.040	0.001	-32.603	0.000
## 61	CLASS	~~	HISTORY	-0.073	0.001	-68.442	0.000
## 62	CLASS	~~	BLOOD	-0.119	0.001	-83.930	0.000
## 63	HISTORY	~~	BLOOD	0.146	0.002	92.212	0.000
## 64	CholCheck	~~	CholCheck	1.000	0.000	NA	NA
## 65	CholCheck	~~	BMI	-0.040	0.000	NA	NA
## 66	CholCheck	~~	Smoker	-0.010	0.000	NA	NA
## 67	CholCheck	~~	PhysActivity	0.004	0.000	NA	NA
## 68	CholCheck	~~	Age	0.090	0.000	NA	NA
## 69	CholCheck	~~	Sex	-0.022	0.000	NA	NA
## 70	CholCheck	~~	HvyAlcoholConsump	-0.024	0.000	NA	NA
## 71	BMI	~~	BMI	1.000	0.000	NA	NA
## 72	BMI	~~	Smoker	-0.017	0.000	NA	NA
## 73	BMI	~~	PhysActivity	0.138	0.000	NA	NA
## 74	BMI	~~	Age	0.016	0.000	NA	NA
## 75	BMI	~~	Sex	-0.091	0.000	NA	NA
## 76	BMI	~~	HvyAlcoholConsump	0.051	0.000	NA	NA
## 77	Smoker	~~	Smoker	1.000	0.000	NA	NA
## 78	Smoker	~~	PhysActivity	-0.087	0.000	NA	NA
## 79	Smoker	~~	Age	0.121	0.000	NA	NA
## 80	Smoker	~~	Sex	0.094	0.000	NA	NA
## 81	Smoker	~~	HvyAlcoholConsump	0.102	0.000	NA	NA
## 82	PhysActivity	~~	PhysActivity	1.000	0.000	NA	NA
## 83	PhysActivity	~~	Age	-0.093	0.000	NA	NA
## 84	PhysActivity	~~	Sex	0.032	0.000	NA	NA
## 85	PhysActivity	~~	HvyAlcoholConsump	0.012	0.000	NA	NA
## 86	Age	~~	Age	1.000	0.000	NA	NA
## 87	Age	~~	Sex	-0.027	0.000	NA	NA

## 88	Age	~~	HvyAlcoholConsump	-0.035	0.000	NA	NA
## 89	Sex	~~	Sex	1.000	0.000	NA	NA
## 90	Sex	~~	HvyAlcoholConsump	0.006	0.000	NA	NA
## 91	HvyAlcoholConsump	~~	HvyAlcoholConsump	1.000	0.000	NA	NA
##	ci.lower	ci.upper					
## 1	1.000	1.000					
## 2	0.924	0.937					
## 3	0.558	0.570					
## 4	0.870	0.883					
## 5	1.000	1.000					
## 6	-0.654	-0.618					
## 7	1.000	1.000					
## 8	1.496	1.606					
## 9	1.000	1.000					
## 10	1.352	1.385					
## 11	1.000	1.000					
## 12	1.415	1.471					
## 13	1.000	1.000					
## 14	0.725	0.748					
## 15	0.146	0.168					
## 16	-0.019	0.004					
## 17	-0.011	0.019					
## 18	-0.045	-0.021					
## 19	0.130	0.189					
## 20	0.339	0.371					
## 21	0.021	0.028					
## 22	-0.141	-0.134					
## 23	-0.012	-0.004					
## 24	-0.014	-0.007					
## 25	0.064	0.071					
## 26	0.017	0.025					
## 27	-0.039	-0.032					
## 28	0.466	0.474					
## 29	0.537	0.545					
## 30	0.827	0.836					
## 31	0.589	0.597					
## 32	0.623	0.645					
## 33	0.846	0.858					
## 34	0.829	0.843					
## 35	0.591	0.620					
## 36	0.670	0.680					
## 37	0.384	0.398					
## 38	0.854	0.865					
## 39	0.700	0.714					
## 40	0.588	0.602					
## 41	0.775	0.786					
## 42	0.804	0.814					
## 43	0.524	0.536					
## 44	0.354	0.377					
## 45	0.157	0.171					
## 46	0.320	0.330					
## 47	0.137	0.145					
## 48	0.397	0.412					
## 49	0.133	0.139					

## 50	-0.069	-0.064
## 51	-0.227	-0.220
## 52	0.155	0.161
## 53	0.232	0.239
## 54	-0.030	-0.026
## 55	-0.142	-0.136
## 56	0.013	0.017
## 57	-0.006	0.001
## 58	0.071	0.076
## 59	-0.020	-0.017
## 60	-0.042	-0.037
## 61	-0.075	-0.071
## 62	-0.122	-0.116
## 63	0.143	0.149
## 64	1.000	1.000
## 65	-0.040	-0.040
## 66	-0.010	-0.010
## 67	0.004	0.004
## 68	0.090	0.090
## 69	-0.022	-0.022
## 70	-0.024	-0.024
## 71	1.000	1.000
## 72	-0.017	-0.017
## 73	0.138	0.138
## 74	0.016	0.016
## 75	-0.091	-0.091
## 76	0.051	0.051
## 77	1.000	1.000
## 78	-0.087	-0.087
## 79	0.121	0.121
## 80	0.094	0.094
## 81	0.102	0.102
## 82	1.000	1.000
## 83	-0.093	-0.093
## 84	0.032	0.032
## 85	0.012	0.012
## 86	1.000	1.000
## 87	-0.027	-0.027
## 88	-0.035	-0.035
## 89	1.000	1.000
## 90	0.006	0.006
## 91	1.000	1.000

#identify degrees of freedom = data points - parameters

#We have 6 multidimensional factors, HEALTH, HEALTHCARE, DIET, CLASS, HISTORY, and BLOOD. GenHlth, PhysHlth, MentHlth, and DiffWalk are all indicators of HEALTH. GenHlth was fixed so it has a standard error of 0. PhysHlth, MentHlth, and DiffWalk are all significant indicators of HEALTH and have a pvalue of 0. NoDocbcCost and AnyHealthcare are the indicators of HEALTHCARE. NoDocbcCost was fixed to 1 and AnyHealthcare is shown to be a significant indicator of HEALTHCARE.

INTERPRETATIONS OF INDICATORS- not sure what this means yet.

#Questions #do we calculate # or parameters based on the complete model or the more simple model?
 #Our EFA analysis worked best with 9 factors but 3 of them only have 1 variable load significantly on it so

we then decide to retain them as response variables right? #The indicators of multidimensional factors all have 1 which is fixed, how do we interpret the significance of this variable on the factor?

#How do we interpret scaled coefficients??