

Eigenvalues of the Covariance Matrix				
	Eigenvalue	Difference	Proportion	Cumulative
1	9.76991794	5.55924638	0.2383	0.2383
2	4.21067156	1.10967667	0.1027	0.3410
3	3.10099489	0.21088690	0.0756	0.4166
4	2.89010799	0.33566625	0.0705	0.4871
5	2.55444173	0.42121307	0.0623	0.5494
6	2.13322866	0.48318506	0.0520	0.6014
7	1.65004360	0.11092305	0.0402	0.6417
8	1.53912055	0.21407287	0.0375	0.6792
9	1.32504768	0.03618664	0.0323	0.7116
10	1.28886104	0.10217738	0.0314	0.7430
11	1.18668367	0.10337894	0.0289	0.7719
12	1.08330473	0.22083611	0.0264	0.7984
13	0.86246862	0.04308193	0.0210	0.8194

Figure 3.1

Table 4.1

Percent Variation Accounted for by Principal Components				
Number of Extracted Factors	Model Effects		Dependent Variables	
	Current	Total	Current	Total
1	24.3127	24.3127	17.9289	17.9289
2	9.3944	33.7072	16.8741	34.8030
3	7.5689	41.2761	4.5546	39.3577
4	6.9016	48.1777	8.3076	47.6653
5	6.2652	54.4428	21.2193	68.8845
6	5.2087	59.6515	4.1018	72.9863
7	4.1137	63.7652	2.4089	75.3953
8	3.8467	67.6119	0.4374	75.8327
9	3.3105	70.9224	0.6715	76.5042
10	3.1992	74.1216	0.6043	77.1085
11	2.9555	77.0771	0.2489	77.3574
12	2.6252	79.7023	2.0086	79.3659

Table 4.2

Parameter Estimates for Centered and Scaled Data		
	x15	x44
Intercept	0.000000 0000	0.000000000 0
x3	-.013399 2067	-.117226583 1
x4	0.004091 7839	-.017452397 2
x5	-.065328 2590	-.027423148 7
x6	-.040579 6543	-.002321978 6
x7	0.019176 1525	-.199763660 6
x8	0.072304 3281	0.016672606 2
x9	0.000977 7501	-.047962977 9
x10	-.039081 0463	-.046935852 1
x11	-.070961 1284	0.000499424 6
x12	0.021175 1531	-.055368381 2
x13	-.007232 9657	0.053892192 0
x14	0.012105 7973	-.152347254 7
x16	-.002963 9021	0.062134694 3
x17	0.049091 6798	0.043301334 9
x18	0.007144 2005	-.080432398 3

x19	0.285891 1237	0.001997316 7
x20	0.273818 3666	0.027068363 4
x21	0.016090 8705	-.047633708 9
x22	-.055090 9090	0.077914616 5
x23	-.039953 8017	-.002764233 8
x24	0.057420 3466	0.121761300 6
x25	0.201481 6594	-.029555135 7
x26	0.030561 7169	-.064320555 9
x27	0.016758 1371	0.097524539 8
x28	0.013292 2817	0.020499682 0
x29	0.018748 5884	0.106961634 9
x30	0.001914 9552	0.056125663 6
x31	-.018326 8147	0.037762620 0
x32	0.032308 5089	0.003041506 0
x33	0.066366 1901	0.061772125 5
x34	-.022834 5823	0.259782545 0
x35	0.174583 8899	0.032791466 5
x36	-.037576 5866	0.011197419 5
x37	0.017198 2052	-.001644595 1

x38	0.057170 4983	-.060565334 1
x39	0.059399 7374	-.048877264 9
x40	-.047272 1227	-.020700615 9
x41	-.080230 1394	-.001366067 4
x42	-.053817 9469	0.030323893 2
x43	0.037576 5866	-.011197419 5

10	CL11	CL29	88	9.1723	
9	CL13	CL20	20	9.4946	
8	CL10	CL9	108	9.7884	
7	CL23	Singapore	3	11.888	
6	CL8	Nigeria	109	12.735	
5	CL6	CL7	112	13.362	
4	China	United States	2	13.717	
3	CL5	India	113	13.864	
2	CL3	Sierra Leone	114	15.363	
1	CL2	CL4	116	16.263	

Figure 5.1

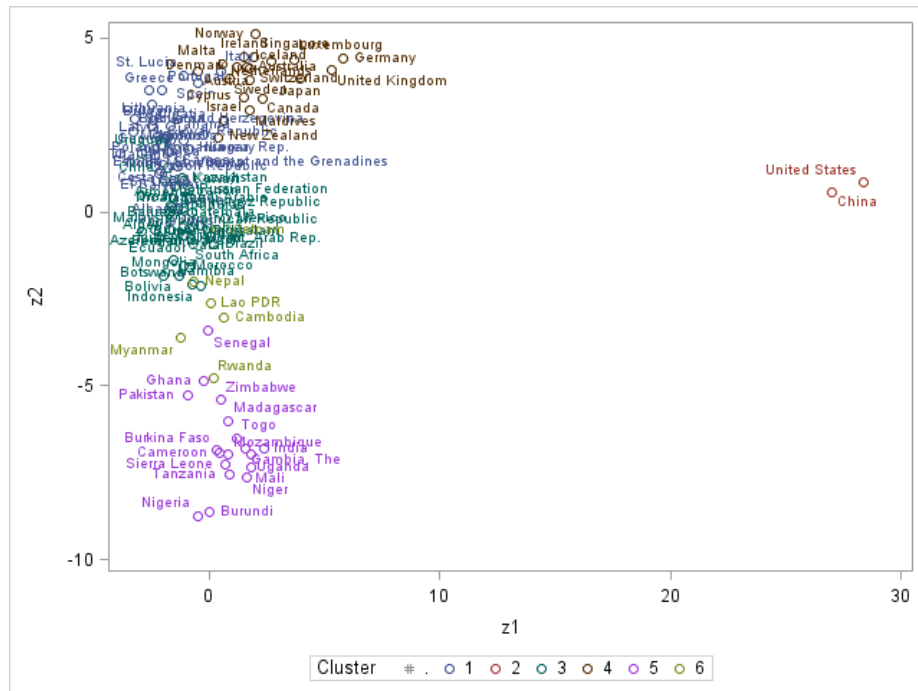


Figure 5.2

Cluster Summary						
Cluster	Frequency	RMS Std Deviation	Maximum Distance from Seed to Observation	Radius Exceeded	Nearest Cluster	Distance Between Cluster Centroids
1	20	0.6811	8.4301	> Radius	3	3.5892
2	2	1.5815	7.2473	> Radius	1	11.8690
3	26	0.6993	6.7748	> Radius	1	3.5892
4	5	1.0833	10.1970	> Radius	5	6.3849
5	41	0.7482	11.5716	> Radius	3	4.0117
6	17	0.9485	10.9857	> Radius	1	5.8078
7	5	1.0476	10.3290	> Radius	3	6.5338

Figure 5.3

Table 6.2

	x1	Score ▼
1	Luxembourg	35.35628284
2	China	33.711449964
3	Qatar	28.700774316
4	Bahrain	25.350123823
5	Iceland	22.662570447
6	United States	20.45187946

Code:

Data standardize and PCA:

```
**Import data;
PROC IMPORT datafile = 'Countries data.xlsx'
    OUT = countries
    DBMS = excel
    replace
    ;
RUN;

**standardize data;
PROC STANDARD DATA=countries MEAN=0 STD=1 OUT=countries;
    VAR x3-x44 ;
RUN;

**print out data to ensure accurate steps;
proc print data = countries;
run;

**PCA;
proc princomp cov;
    var x3-x44;
run;
```

Multivariate PCA regression

```
proc standard data = work.country mean = 0 std =1 out = countries;
    var x3-x44;
run;
proc pls data = countries method = PCR nfac=12;
model x15 x44 = x3-x15 x16-x43/solution;
run;
```

Cluster analysis

```
RUN;

** 2.1 Hierarchical Clustering - Average Linkage;

proc cluster data=countries outtree=treecountry method=average nonorm;

    var x3-x44;

    id country;

run;

* 2.2 Hierarchical Clustering - Centroid;

proc cluster data=countries outtree=treecountry method=centroid nonorm;

    var x3-x44;

    id country;

run;

* K-means Clustering;

* 2.3 Principal Component Analysis;

proc princomp data=countries out=countriesPC;

var x3-x44;

run;

goptions reset=all;

symbol pointlabel=("#country") value=dot;

proc gplot data=countriesPC;

plot prin2*prin1/ vaxis=axis2 haxis=axis1 nolegend;

axis1 label=("z1" justify=center);

axis2 label=("z2" justify=center a=90);

run;

* 2.4 K-means - Random 6 observations;

proc fastclus data=countries maxc=6 replace=random maxiter=10
out=Clus_out1 radius=1;

var x3-x44;

id country;

run;

proc sort data=Clus_out1;
```

```

by cluster distance;

run;

proc print data=Clus_out1;
var country cluster distance;

run;

proc candisc data=Clus_out1 noprint out=ProCan1(keep=country cluster
Can1 Can2);

class cluster;

var x3-x44;

run;

proc sgplot data=ProCan1;

scatter y=Can2 x=Can1 / group=cluster datalabel=country;

label Can1="z1" Can2="z2";

run;

* 2.5 K-menas - First 7 observations;

proc fastclus data=countries radius=1.5 maxc=7 replace=none maxiter=10
out=Clus_out2;

var x3-x44;

id country;

run;

proc sort data=Clus_out2;

by cluster distance;

run;

proc candisc data=Clus_out2 noprint out=ProCan2(keep=country cluster
Can1 Can2);

class cluster;

var x3-x44;

run;

proc print data=Clus_out2;

```

```

var country cluster distance;

run;

proc sgplot data=ProCan2;
scatter y=Can2 x=Can1 / group=cluster datalabel=country;
label Can1="z1" Can2="z2";
run;

/* Method # for getting seeds: Use Average Linkage to get cluster
centriods */

proc cluster data=countries method=average outtree=CunTree noprint
nonorm;

var x3-x44;

id country;

run;

proc tree data=CunTree nclusters=6 out=newdata noprint;

id country;

copy x3-x44;

run;

proc sort data=newdata;

by cluster;

run;

proc print data=newdata;

var country cluster;

run;

proc means data=newdata;

by cluster;

output out=Seeds mean=x3-x44;

var x3-x44;

run;

proc fastclus data=countries maxc=7 maxiter=50 seed=Seeds
out=Clus_out4;

var x3-x44;

id country;

```

```

run;

proc sort data=Clus_out4;
by cluster distance;
run;

proc print data=Clus_out4;
var country cluster distance;
run;

proc candisc data=Clus_out4 out=ProCan4(keep=country cluster Can1 Can2)
noprint;

class cluster;

var x3-x44;

run;

proc sgplot data=ProCan4;
scatter y=Can2 x=Can1 / group=cluster datalabel=country;
label Can1="z1" Can2="z2";
run;

```

FA

```

PROC FACTOR DATA=countries ROTATE=VARIMAX OUTSTAT=SCORES SCORE ALL;
VAR X3-X44;
RUN;

PROC SCORE DATA=countries SCORE=SCORES OUT=FINALSCORE;
VAR X3-X44;
RUN;

```