

Session 9-10, Dimensionality Reduction and Derived Attributes (Technical Slides)

T. Evgeniou, A. Ovchinnikov, INSEAD

What is Dimensionality Reduction Factor Analysis?

Derive new variables which are (linear) combinations of original ones and capture most of the information in the data.

Is often used as a first step in Data Analysis

Can also be used to solve multicollinearity issues

Factor Analysis: Key ideas

1. Transform the original selected variables into a few factors
2. Understand the underlying structure of the data in terms of factors
3. Use the factors for subsequent analysis

Key Questions

1. Can we really simplify the data by grouping the attributes?
2. How many factors should we use?
3. How good are the factors we found?
4. How interpretable and actionable are the factors?

Dimensionality Reduction and Analysis: 6 (Easy) Steps

1. Confirm data is metric
2. Scale the data
3. Check correlations
4. Choose number of factors
5. Interpret the factors
6. Save factor scores

Applying Factor Analysis: Evaluation Applications

Variables available:

- GPA
- GMAT score
- Scholarships, fellowships won
- Evidence of Communications skill
- Prior Job Experience
- Organizational Experience
- Other extra curricular achievement

Which variables are correlated? What do these capture?

Example Factors

	Variables	Component 1	Component 2
1	GPA	0.96	-0.05
2	GMAT	0.95	0.19
3	Fellow	0.95	-0.01
4	Comm	0.7	0.54
5	Job.Ex	0.19	0.93
6	Organze	0.01	0.89
7	Extra	0.01	0.86

Step 1: Confirm data is me

	Variables	GPA	GMAT	Fellow	Comm	Job.Ex	Organ
1	1	3	580	2	3.5	5	3.8
2	2	3.2	570	2	3.8	6	3.8
3	3	3.7	690	3	3.3	3	3.2
4	4	3.9	760	3	3.8	5	3.9
5	5	2.8	480	2	3.2	6	3.8
6	6	3.4	520	2.5	2.6	2	2.5
7	7	3.6	670	3	3.7	4	3.5
8	8	3.6	760	3	3.9	5	3.3

Step 2: Scale the data

	Variables	min	X25.percent	median	mean	X75.percent	max
1	GPA	2.5	2.8	3.45	3.31	3.62	3.9
2	GMAT	380	480	575	583.5	682.5	730
3	Fellow	1	2	2.8	2.45	3	3.9
4	Comm	2	3.18	3.4	3.34	3.73	3.9
5	Job.Ex	2	3	5	4.25	5.25	6
6	Organze	1	3.05	3.4	3.2	3.8	3.9
7	Extra	2.4	2.88	3.4	3.3	3.8	4

Data Standardization: Example

```
ProjectDatafactor_scaled=apply(ProjectDatafactor, 2,
function(r) {
  if (sd(r)!=0) {
    res=(r-mean(r))/sd(r)
  } else {
    res=0*r; res
  }
})
```

Standardized Data: Summary S

	Variables	min	X25.percent	median	mean	X75.percent
1	GPA	-1.72	-1.08	0.31	0	0.68
2	GMAT	-1.7	-0.87	-0.07	0	0.83
3	Fellow	-1.6	-0.5	0.39	0	0.61
4	Comm	-2.73	-0.33	0.13	0	0.8
5	Job.Ex	-1.48	-0.82	0.49	0	0.66
6	Organze	-2.99	-0.2	0.27	0	0.82
7	Extra	-1.75	-0.83	0.19	0	0.97

Step 3: Check correlation

GPA	GMAT	Fellow	Comm	Job.Ex	Organze
1	0.9	0.92	0.56	0.15	-0.03
0.9	1	0.86	0.78	0.33	0.19
0.92	0.86	1	0.59	0.18	0.01
0.56	0.78	0.59	1	0.6	0.47

Step 4. Choose number of factors

For the method considered here (Principal Component Analysis):

- If there are n variables we will have n factors in total
- First factor will explain most of the variance, second factor will explain less, and so on.

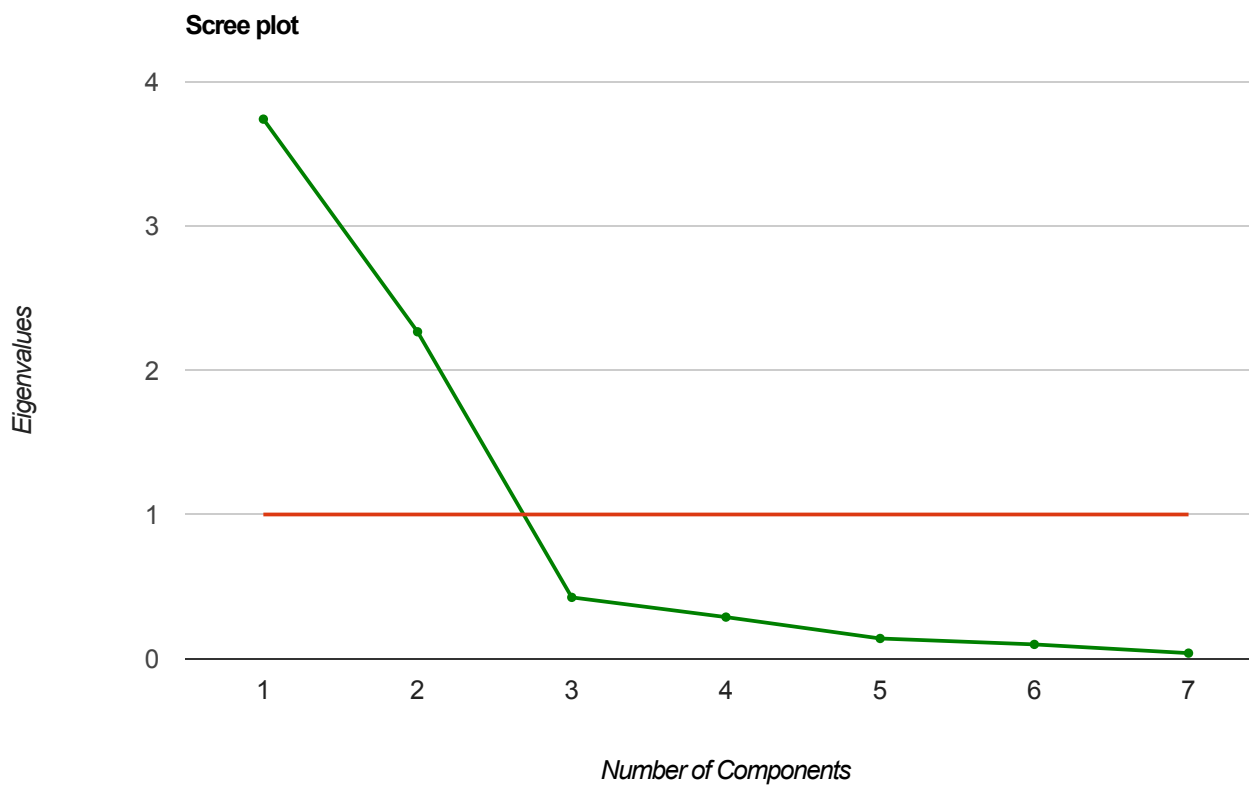
Eigenvalues and Variance Explained by Factors

- each factor will have an associated eigenvalue corresponds to the amount of variance explained by each factor
- with standardized variables each variable has a variance of 1 and the sum of all eigenvalues with n raw attributes is n
- we would like to capture as much of the total variance as possible, while keeping as few factors as possible

How Many Factors? Eigenvalues ance Explained

Components	Eigenvalue	Percentage_of_explained_variance	Cumulative_percentage_of_explained_variance
Component No:1	3.74	53.48	53.48
Component No:2	2.27	32.4	85.88
Component No:3	0.42	6.07	91.95
Component No:4	0.29	4.11	96.06
Component No:5	0.14	1.99	98.05
Component No:6	0.1	1.41	99.46
Component No:7	0.04	0.54	100

How Many Factors? Scree



How many factors?

Three criteria to use:

- Eigenvalue > 1
- Cumulative variance explained
- “Elbow” in the Scree plot

Using the eigenvalue criterion we select 2

Step 5. Interpret the factors

Rotated Selected Factors using the varimax

	Variables	Component 1	Component 2
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4	Comm	0.7	0.54
5	Job.Ex	0.19	0.93
6	Organze	0.01	0.89
7	Extra	0.01	0.86

For visualization, let's suppress the numbers...

	Variables	Component 1	Component 2
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What Factor Loads "Look Good"

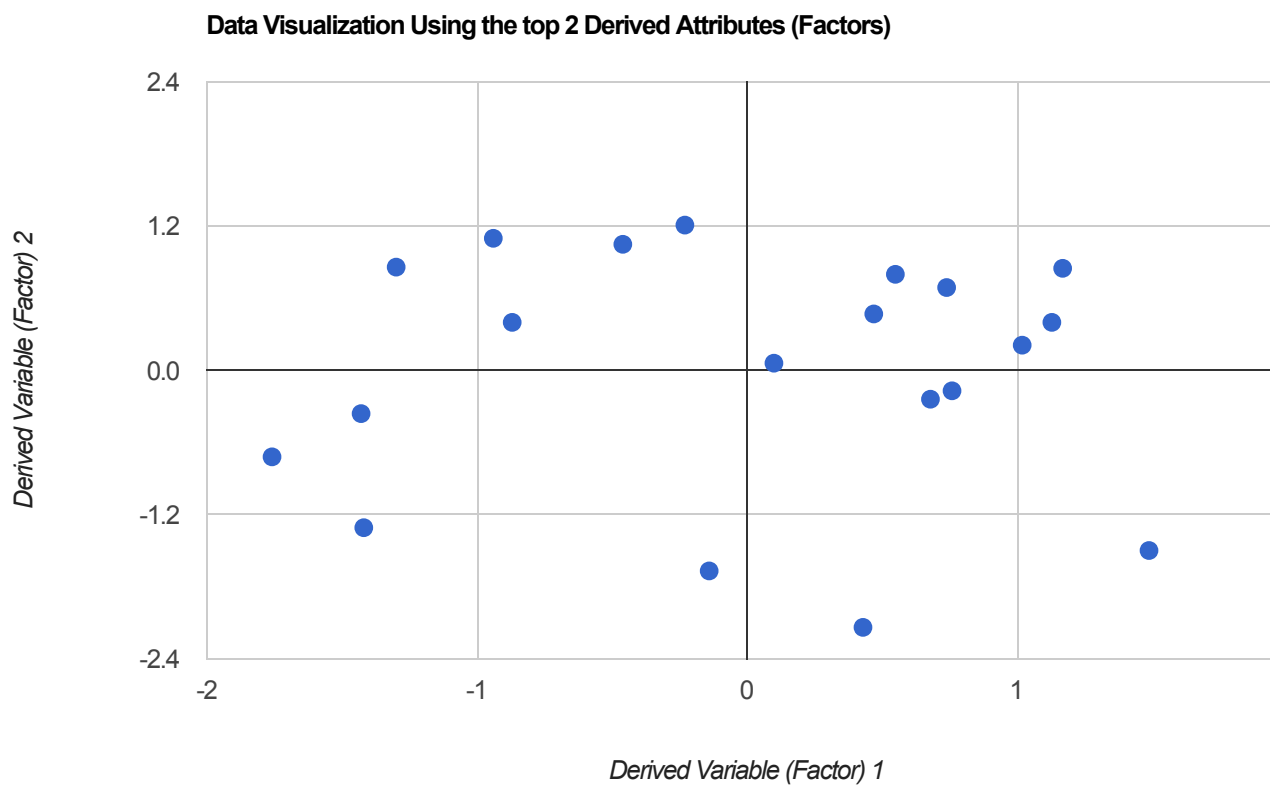
Technical Quality Criteria

1. For each factor (column) only a few loadings are high (large absolute value)
2. For each raw attribute (row) only a few loadings are high (large absolute value)
3. Any pair of factors (columns) should have different patterns of loading

Step 6. Save factor scores

	Observation	Derived Variable (Factor) 1	Derived Variable (Factor)
1	1	-0.46	1.05
2	2	-0.23	1.21
3	3	0.68	-0.24
4	4	1.13	0.4
5	5	-0.94	1.1
6	6	-0.14	-1.67
7	7	0.76	-0.17
8	8	1.02	0.21

Using the Factor Scores: Percep



Factor Analysis: Some (Tech Concepts

1. Correlation
2. Variance explained (eigenvalues)
3. Scree plot
4. varimax rotation
5. Factor Loadings (“components”)
6. Factor scores

Key Questions

1. How many factors should we use? Why? Quantitative criteria
Qualitative criteria
2. How can we name and interpret the factors?
3. What are some issues to consider?