



























### Main steps after generating your item pool

- 1. Write the introduction to your questionnaire
- 2. Evaluate the content of each item
- 3. Evaluate Inability or unwillingness to answer

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- 4. Critically asses question wording
- 5. Determine the order of questions
- 6. Determine form and layout
- 7. Determine method of administring
- 8. Pretest your questionnaire
- 9. Iterate

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#### Identify unwillingness or inability to answer

1. Minimize the effort required of participants

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- 2. Is the context of the questions clear and appropriate
- 3. Make a request for information seem legitimate
- 4. In case of sensitive information:
  - 1. Place items at the end
  - 2. Preface with a 'common' statment
  - 3. Ask questions in 3rd person
  - 4. Hide questions in between others
  - 5. Provide response categories
  - 6. Use randomized techniques

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Bad examples 5: Jarg	gon
Bad	Better
Using the system made me feel more socially connected to my social network: Agree Disagree	I felt closer to my friends by using the system:DisagreeAgree000000
/ Industrial Design	TU/e Technische Universiteit Eindhoven University of Technology B/142000 PAGE 24







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Ans	W	er			le	go	nes	
What is you	ır over	allsa	atista	action	with	ourproc	luct?	<ul> <li>For likert type scales:</li> </ul>
Not at all satisfied	0	0	0	0	0	Extre	emely fled	Even or odd?
What is you	ir over	all sa	atisfa	ction	with	our proc	duct?	• Labeling?
Not at all satisfied	1 ©	2 ()	3 ()	4 0	5 0	Extre satist	mely fled	<ul> <li>Number of options?</li> <li>"Don't know"?</li> </ul>
What is you	r over	all sa	atisfa	ction	with	our prod	juct?	Pictorials?
0 1	02	0	3 (	04	05			
What is you	r over	all sa	atisfa	ction	with	our prod	luct?	
Not at all satisfied	Sligh satisf	tly i ied	Mod sat	erate isfied	ly si	Very atisfied	Extremely satisfied	
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6	item19	Numeri	Correlate	Se Elvariate     to us	a The system was ea
7	item19	Numeri	Burneroim	E. Defid	db Hontoming the backs
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10	item22	Numeria	Classify	e system nerformed as 1	Correlation Coefficients
11	item73	Numeria	Dimension Reduction	b a input were exectly und	Paarton Kandalit tau b Spaarnan
12	item24	Numeria	Scale	e interaction with the sys:	Test of Significance
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18			Mutple Imputation	•	
19			Complex Samples	•	
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1	var1 Numeric	0 2 The system was easy to use	None No	ne O	≣ Right @ Scale
2	var2 Numeric	8 2 I had problems using the system	None No	ne O	TRight & Scale
з	var3 Numeric	8 2 Using the system was simple	None No	ne 8	I Right 🖉 Scale
4	var4 Numeric	8 2 Lenjoyed using the system	None No	ne 8	≡ Right 🖋 Scale
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	Descriptive Statistics	Label			
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Scale reliability	
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/ Industrial Design	TU/e Technische Universiteit Endioven University of Technology 16.0.2000 PAGE 50

Scale r	reliability	correlati	on matrix	(						
	Inter-Item Correlation Matrix									
	Var 1	Var 2	Var 3	Var 3						
Var 1	1,000	-,565	,631	,670						
Var 2	-,565	1,000	-,807	-,678						
Var 3	,631	-,807	1,000	,662						
Var 3	,670	-,678	,662	1,000						
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Scale	e re	eliat	oilit	y SP	SS ou	tput		
Re	liability \$	Statistics						
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a. The valu covarian assumpt								
Γ	;	Scale Me Item Del	ean if eted	Scale Variance if Item Deleted	Corrected Item-Total Correlation	Squared Multiple Correlation	Cronbach's Alpha if Item Deleted	
	/ar 1	12,	1667	4,144	,641	,512	-1,747 <sup>a</sup>	-
1	/ar 2	13,	1000	20,093	-,786	,689	,847	
\ \	/ar 3	12,	7000	4,769	,329	,701	-1,036 <sup>a</sup>	
\ \	/ar 3	13,	2333	4,668	,551	,585	-1,410 <sup>a</sup>	
	a. The viol	e value is ates relia	negative bility mo	e due to a nega del assumption	ative average co ns. You may wa	ovariance among nt to check item	titems. This codings.	echnische Universiteit Indhoven Iniversity of Technology
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	Inter-Iten	n Correlatior	n Matrix	
	Var 1	Var 3	Var 4	Var 2 Mirror
Var 1	1,000	,631	,670	,565
Var 3	,631	1,000	,662	,807
Var 4	,670	,662	1,000	,678
Var 2 Mirror	,565	.807	,678	1,000

ite	liability Statistics				
Cronbach's Alpha ,889	Cronbach's Alpha Based on Standardized Items ,890	N of Items 4			
		Item-Total	Statistics		
	Scale Mean if Item Deleted	Item-Total Scale Variance if Item Deleted	Statistics Corrected Item-Total Correlation	Squared Multiple Correlation	Cronbach's Alpha if Item Deleted
Var 1	Scale Mean if Item Deleted 12,2333	Item-Total Scale Variance if Item Deleted 21,771	Statistics Corrected Item-Total Correlation ,689	Squared Multiple Correlation ,512	Cronbach's Alpha if Item Deleted ,882
Var 1 Var 3	Scale Mean if Item Deleted 12,2333 12,7667	Item-Total Scale Variance if Item Deleted 21,771 18,323	Statistics Corrected Item-Total Correlation ,689 ,806	Squared Multiple Correlation ,512 ,701	Cronbach's Alpha if Item Deleted ,882 ,840
Var 1 Var 3 Var 4	Scale Mean if Item Deleted 12,2333 12,7667 13,3000	Item-Total Scale Variance if Item Deleted 21,771 18,323 21,183	Statistics Corrected Item-Total Correlation ,689 ,806 ,757	Squared Multiple Correlation ,512 ,701 ,585	Cronbach's Alpha if Item Deleted ,882 ,840 ,859













### Steps to perfoming factor analysis Determine items Use your generated item set Get sufficient N Minimum 100 Minimum 300

- 50 + 5\*m
- Determine number of factors
  - Scree plot, Eigenvalues, Explained variance
- Rotate the factor solution for a simpler structure
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- Compute factor scores
  - Regression

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684			Regota	• <u>• • • • • • •</u>	
	Name	Ту	Dgueriptive Statistics	Labe	
1	item13	Numeri	Tugerz	system looked good	
2	item14	Numerio	Compute Means	<ul> <li>le design of the system was pleasin;</li> </ul>	
3	item15	Numerio	General Linear Model	<ul> <li>ked the look and feel of the system</li> </ul>	
4	item16	Numeria	Generalized Linear Models	In design of the icons was good	
5	item17	Numerio	Miged Models	Ie system looked beautiful	
6	item18	Numeric	Correlate	e system was easy to use	
/	item19	Numeric	Regression	<ul> <li>fould easily perform the tasks I want</li> </ul>	
8	item.20	Numen	Logineer	e system was easy to work with	
9	item/21	Numeri	Neural Networks	bitoming the tasks was easy	
10	item22	Numeric	Classely	je system performed as l'expected	
11	item23	Numeric	Unension Heauction	In Econor	
14	Demi24	reament	biogen	Ch. Cation Scales	
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10	-		Forecasping Sumbusi		
			Multiple Response		
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PAF SPSS 3	
Right       Right       Nominal         Set Factor Analysis: Extraction       Image: Comparison of the compa	Image: Sector Analysis: Rotation         Method         Bone       Quartmax         Yarimax       Equamax         Display       gromax         Display       Gotated solution         Maginum Iterations for Convergence:       25         Continue       Cancel         Help       Help
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KMO	and Bartle	tt's t	test
	KMO and Bartlett's Test		<ul> <li>KMO</li> <li>Index magnitude of observed correlations versus partial correlations</li> </ul>
Kaiser-Meyer-Olk	in Measure of Sampling Adequacy.	,935	
Bartlett's Test of Sphericity	Approx. Chi-Square df Sig.	8890,014 66 ,000	<ul> <li>Barlett's</li> <li>Test of correlation of the variables</li> <li>Should be significant: Sig. &lt; 0.05</li> </ul>
/ Industrial Design			TU/e Technische Universitieit Eindhoven University of Technology Bit M2000 PAGE 15

Communalities						
	Initial	Extraction				
The system looked good	,566	,665				
The design of the system was pleasing	,566	,626				
I liked the look and feel of the system	,555	,600				
The design of the icons was good	,433	,444				
The system looked beautiful	,582	,619				
The system was easy to use	,570	,663				
I could easily perform the tasks I wanted to perform	,460	,544				
The system was easy to work with	,330	,366				
Performing the tasks was easy	,511	,550				
The system performed as I expected	,400	,452				
The icons were easy to understand	,661	,644				
The interaction with the system was pleasant	,528	,466				

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Factor         Total         % of Variance         Cumulates         Estraction         Sums of Squared Loading         Loading2           1         0.271         52.260         52.260         5.839         48.655         5.721           2         1.234         10.283         62.743         9.00         6.669         55.324         4.815           3         .711         5.926         68.469         6.669         55.324         4.815           4         .608         5.066         73.535         6         4.697         73.351         6         4.991         4.155         82.2871         6         4.991         4.155         82.3871         6         4.991         4.92.571         10         .329         2.745         95.316         4         9         3.074         92.571         100.000         1         1         10         .329         2.745         95.316         1         1         100.000         1         1         100.000         1         1         100.000         1         1         1         100.000         1         1         100.000         1         1         1         100.000         1         1         100.000         1         1         10	Initial Eigenvalues         Extraction Sums of Squared Loadings         Loadings           Factor         Total         % of Variance         Cumulative %         Total           1         0.271         \$\$22260         \$\$2280         \$\$8 of Variance         Cumulative %         Total           2         1.234         10.283         62.243         \$\$00         6\$.669         55.324         4.81           4         .608         5.066         73.535         6         6.469         55.324         4.81           5         .564         4.697         73.535         6         4.697         73.334         6         9.0497         9         3.394         3.106         9.0497         9         3.393         3.074         92.571         10         10         3.39         2.745         95.316         10         11         3.39         2.745         95.316         10         11         3.39         2.745         95.316         10         10         10         3.39         2.710         10.0000         10         12         2.33         2.110         10         10.0000         12         2.33         2.110         10.0000         12         2.33         2.110         10.0000         12 </th <th></th> <th></th> <th></th> <th>Total Var</th> <th>iance Explain</th> <th>ied</th> <th> </th> <th>Rotation Sums of</th>				Total Var	iance Explain	ied		Rotation Sums of	
Factor         Total         % of Variance         Cumulative %         48.653         4.813           4         .608         5.066         73.353            48.631 <th><math display="block">\begin{array}{r rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr</math></th> <th></th> <th></th> <th>Initial Eigenvalu</th> <th>ies</th> <th>Extraction</th> <th>Sums of Square</th> <th>ed Loadings</th> <th colspan="2">Loadings<sup>8</sup></th>	$\begin{array}{r rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$			Initial Eigenvalu	ies	Extraction	Sums of Square	ed Loadings	Loadings <sup>8</sup>	
1         6.271         52.260         5.280         5.839         48.655         48.655         5.121           2         1.234         10.233         62.250         5.839         48.655         48.655         5.121           3         7.711         5.926         68.469         6.669         55.324         4.813           4         .608         5.066         7.3535         6         4.697         76.322         6           6         .499         4.155         82.387         6         6.949         9         3.944         3.916         80.497         9         3.384         3.196         69.497         9         3.022         2.745         95.316         1         1         1.302         2.745         95.316         1         1         1         1.302         2.745         95.316         1         1         1         1.10         10.0000         1         1         1         1.10         1.10         10.0000         1         1         1         1.10         1.10         10.0000         1         1         1         1.10         1.10         1.10         1.10         1.10         1.10         1.10         1.10         1.10         1.10<	$\begin{array}{c c c c c c c c c c c c c c c c c c c $	Factor	Total	% of Variance	Cumulative %	Total	% of Variance	Cumulative %	Total	
2         1.234         10.283         62.543         ,800         6.669         55.324         4.81:           3         7.11         5.926         68.469         68.469         55.324         4.81:           4         .608         5.066         73.555         5         5.664         4.897         78.232         68.409         68.409         69.669         55.324         4.81:           5         .564         4.897         78.232         6.669         5.024         4.81:           8         .344         .108         68.407         6.669         5.024         4.81:           9         .369         3.074         92.571         6.5316         6.669         5.5324         4.81:           10         .329         2.745         95.316         6.699         5.5324         4.81:           11         .309         2.574         97.880         6.699         5.5324         4.81:           e.tractor Method: Principal Avis Factoring.         a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	1	6,271	52,260	52,260	5,839	48,655	48,655	5,126	
3       .711       5.926       68,469         4       .608       50.66       73,535         5       .564       4.897       78,232         6       .499       4,155       62,367         7       .470       .3914       86,301         8       .384       .3196       89,407         9       .369       .3074       92,571         10       .329       2,745       95,316         11       .309       2,574       97,880         12       .253       2,110       100,000         Extraction Method: Principal Asis Factoring:         a. When factors are correlated, sums of squared loadings cannot be added to obtain a total vertance.         Ce in the items explained boy the factors	$\begin{array}{c c c c c c c c c c c c c c c c c c c $	2	1,234	10,283	62,543	,800	6,669	55,324	4,813	
4       .008       5.066       73,535         5       .564       4,697       78,232         6       .499       4,155       82,387         7       .470       3.914       86,301         8       .384       3,196       69,497         9       .369       .3074       92,571         10       .329       .2745       95,316         11       .303       2,574       100,000         12       .253       2,110       100,000         L2       .253       2,110       100,000         .253       .2,110       100,000          Bardeno Method: Principal Asis Factoring.       a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         Ce in the items explained by the factors	$\begin{array}{c c c c c c c c c c c c c c c c c c c $	3	,711	5,926	68,469					
5         .564         4.697         78.232           6         .499         4.155         82.287           7         .470         3.914         66.301           8         .344         3.196         89.497           9         .369         3.074         92.571           10         .329         2.745         95.316           11         .309         2.574         97.890           12         .253         2.110         100.000   Extraction Method. Principal Avis Factoring. a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance. <b>2.6 in the items explained by the factors</b>		4	,608	5,066	73,535					
6       499       4,155       82,387         7       470       3.914       86,301         8       3.84       3.196       89,497         9       3.69       3.074       92,571         10       3.29       2,745       95,316         11       3.09       2,574       97,880         12       2.53       2,110       100,000         Extraction Nethod: Principal Asis Factoring:         a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.	6       .499       .4155       82.387         7       .470       .3914       82.387         8       .384       .3196       89.407         9       .389       .3074       92.571         10       .329       .2745       95.316         11       .309       .2574       97.880         12       .253       .110       100.000         Extraction Method: Principal Abs Factoring:         a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         Ce in the items explained by the factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         Ce in the items explained by the factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         Ce in the items explained by the factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         Ce in the items explained by the factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         Action explained by the factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         Ce in the items explained by the factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         Ce in the items explained by the factors are co	5	,564	4,697	78,232					
7       470       3.914       66.301         8       3.44       3.106       69.497         9       3.68       3.074       92.571         10       3.29       2.745       95.316         11       3.09       2.574       97.880         12       2.53       2.110       100.000         Extraction Method: Principal Avis Factoring.         a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         Ce in the items explained by the factors	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	6	,499	4,155	82,387					
8         .394         3.196         89.497           9         .369         3.074         92.571           10         .329         2.745         95.316           11         .309         2.574         97.890           12         .253         2.110         100.000           Extraction Method: Principal Akis Factoring.         a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.           Ce in the items explained by the factors	$\begin{array}{c c c c c c c c c c c c c c c c c c c $	7	,470	3,914	86,301					
9         .369         3.074         92.571           10         .329         2.745         95.316           11         .309         2.574         97.890           12         .253         2.110         100.000           Extraction Metho: Principal Alse Factoring.           a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.	9       .369       .3.745       92.571         10       .329       .2.745       95.316         11       .303       .2.574       97.800         12       .253       .2.110       100.000         12       .253       .2.110       100.000         14       .309       .2.574       97.800         15       .253       .2.110       100.000         16       .000       .000       .000         17       .253       .2.110       100.000         18       .267       .000       .000         19       .000       .000       .000         12       .253       .110       100.000         12       .253       .2.110       .000         12       .253       .010       .000         12       .253       .010       .000         10       .000       .000       .000         .253       .010       .000       .000         .253       .010       .000       .000         .253       .010       .000       .000         .253       .000%       .000       .000         .254       .000%<	8	,384	3,196	89,497					
10       .329       2.745       95.316         11       .309       2.574       97.880         12       .253       2.110       100.000         Extraction Method: Principal Avis Factoring.         a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         Ce in the items explained by the factors	10       .329       2.754       95.316         11       .309       2.574       97.890         12       .253       2.110       100.000         Extraction Method: Principal Avis Factoring.         a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         ce in the items explained by the factors ative > 60%         hyperbolic colspan="2">hyperbolic colspan="2">hyperbolic colspan="2">hyperbolic colspan="2">hyperbolic colspan="2">hyperbolic colspan="2">a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         ce in the items explained by the factors ative > 60%         hyperbolic colspan="2">hyperbolic colspan="2"         hyperbolic cols	9	,369	3,074	92,571					
11       .309       2.574       97,890         12       .253       2.110       100,000         Extraction Wethod: Principal XMs Factoring.       a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         Ce in the items explained by the factors	11       .209       2.574       97,890         12       .253       2.110       100,000         Extraction Method: Principal Avis Factoring.         a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         Ce in the items explained by the factors ative > 60%         August colspan="2">August colspan="2"         August colspan="2" <td>10</td> <td>,329</td> <td>2,745</td> <td>95,316</td> <td></td> <td></td> <td>   </td> <td></td>	10	,329	2,745	95,316					
12       283       2,110       100,000         Extraction Method: Principal Avis Factoring.       a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         ce in the items explained by the factors	12       253       2.110       100.000         Extraction Method: Principal Avis Fectoring.       a. When factors are correlated, sums of squared loadings cannot be added to obtain a total variance.         ce in the items explained by the factors at ive > 60%         bative > 60%         by value < 1 – Factor explains less variance ual item	11	,309	2,574	97,890					
	ative > 60% nvalue < 1 – Factor explains less variand ual item	extrac a. Wh	ition Method en factors a	1: Principal Axis Fa are correlated, sun <b>e item</b>	ctoring. 1s of squared load	ings cannot b	ne added to obtai	n a total variance.	tors	

#### Communality

 Measures the percent of variance in a given variable explained by all the factors jointly.

- Example:
  - Factor solution explains the system "looking good" very well, but poorly explains the "perfomed as expected" item



Factor ma	trix		
Factor Mat	rixª		<ul> <li>Correlations between</li> </ul>
	Fact	or	town own footowe
The evolution locked aread	1	2	items and factors
The system looked good The design of the system was pleasing	,718	-,388 -,341	<ul> <li>Factor loadings</li> </ul>
I liked the look and feel of the system	,734		Suppressed values
The design of the icons was good	,639		below 0.3
The system looked beautiful	,784		Eactor 1 correlatos
The system was easy to use	,737	,347	<ul> <li>Factor i correlates</li> <li>bighty with all itoms</li> </ul>
I could easily perform the tasks I wanted to perform	,613	,410	nighty with an items
The system was easy to work with	,555		<ul> <li>Solution not simple to</li> </ul>
Performing the tasks was easy	,731		interpret.
The system performed as I expected	,619		<ul> <li>High correlation means</li> </ul>
The icons were easy to understand	,802		highly roprosontative for
The interaction with the system was pleasant	,682		the factor
Extraction Method: Princip	al Axis Facto	ring.	
a. 2 factors extracted. 6 ite	rations requ	ired.	
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Oblimin	rotat	ion	
P	attern Matrix*		<ul> <li>Factor loadings after</li> </ul>
	Fac	tor	i deter ieddinige arter
The system looked	1	2	rotating
The design of the s	ystem ,831		
was pleasing I liked the look and the system	feel of ,731		<ul> <li>Factor one: design of the</li> </ul>
The design of the id was good	cons ,607		system
The system looked beautiful	,555	,302	Factor two: Fase of use
The system was eause	asyto	,794	
I could easily perfor tasks I wanted to p	rm the erform	,810	
The system was ea work with	asy to	,572	<ul> <li>Use highest correlations</li> </ul>
Performing the task easy	ks was	,518	to name the factors
The system perform I expected	ned as	,632	to name the factors
The icons were eas understand	syto ,509	,370	
The interaction with system was please	n the ,397 ant	,352	Correlation factors 0.0
Extraction Metho Rotation Methor Normalization.	od: Principal Axis Facto d: Oblimin with Kaiser	pring.	Correlation factors: 0.6
a. Rotation conv	erged in 10 iterations		
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