# Joachim\_CFA\_report

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# Zaczynam od analizy GPTS

## avevar 0.6288829

0.7011921

```
## This is lavaan 0.6-8
## lavaan is FREE software! Please report any bugs.
##
## This is semTools 0.5-4
## All users of R (or SEM) are invited to submit functions or ideas for functions.
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
     filter, lag
## The following objects are masked from 'package:base':
##
##
     intersect, setdiff, setequal, union
Pierwszy mnodel tak jak w pdfie:
## chisq.scaled
               df.scaled
                          cfi.scaled
                                    tli.scaled rmsea.scaled
                                                               srmr
##
     2761.370
                 134.000
                              0.938
                                         0.929
                                                   0.096
                                                               0.072
## For constructs with categorical indicators, the alpha and the average variance extracted are calcula
        reference persecutory
## alpha 0.9296015
                  0.9548849
## omega 0.8978957
                  0.9321378
## omega2 0.8978957
                  0.9321378
## omega3 0.9030990
                  0.9627104
```

Drugi model z wyrzuconym itemem 8\_A, który ładował dwie skale (usunięcie na podstawie mod indices.)

```
## chisq.scaled df.scaled cfi.scaled tli.scaled rmsea.scaled srmr
## 1780.189 118.000 0.958 0.951 0.081 0.062
## For constructs with categorical indicators, the alpha and the average variance extracted are calcula
## reference persecutory
```

```
## alpha 0.9239770 0.9548849
## omega 0.8859223 0.9319913
## omega2 0.8859223 0.9319913
## omega3 0.8848520 0.9615662
## avevar 0.6391935 0.7005812
```

Różnica w teście robust chi square między modelem 1 i 2

```
## [1] 889.895 0.000
```

model 3 z dodaną kowariancją między part<br/>B\_gptsb2  $\sim$  partB\_gptsb4 na podstawie mod indices:

```
## chisq.scaled df.scaled cfi.scaled tli.scaled rmsea.scaled srmr
## 1634.961 117.000 0.961 0.955 0.078 0.059
```

## For constructs with categorical indicators, the alpha and the average variance extracted are calcula

```
## reference persecutory
## alpha 0.9239770 0.9548849
## omega 0.8859170 0.9256482
## omega2 0.8859170 0.9256482
## omega3 0.8848528 0.9503507
## avevar 0.6392014 0.6932342
```

Różnica w teście robust chi square między modelem 2 i 3

```
## [1] 59.029 0.000
```

model 4 z dodaną kowariancją między part<br/>B\_gptsb3  $\sim$  partB\_gptsb4 na podstawie mod indices:

```
## chisq.scaled df.scaled cfi.scaled tli.scaled rmsea.scaled srmr
## 1526.745 116.000 0.964 0.958 0.076 0.057
```

## For constructs with categorical indicators, the alpha and the average variance extracted are calcula

```
## reference persecutory
## alpha 0.9239770 0.9548849
## omega 0.8859102 0.9186102
## omega2 0.8859102 0.9186102
## omega3 0.8848366 0.9395323
## avevar 0.6392011 0.6871400
```

#### Różnica w teście robust chi square między modelem 3 i 4

```
## [1] 62.986 0.000
```

model 5 z dodaną kowariancją między part<br/>B\_gptsb2  $\sim$  partB\_gptsb3 na podstawie mod indices:

```
## chisq.scaled df.scaled cfi.scaled tli.scaled rmsea.scaled srmr ## 1389.498 115.000 0.968 0.962 0.072 0.055
```

## For constructs with categorical indicators, the alpha and the average variance extracted are calcula

```
## reference persecutory
## alpha 0.9239770 0.9548849
## omega 0.8859040 0.9101319
## omega2 0.8859040 0.9101319
## omega3 0.8848225 0.9268061
## avevar 0.6392005 0.6799656
```

Różnica w teście robust chi square między modelem 4 i 5

```
## [1] 72.04 0.00
```

### Teraz będzie MUSEQ

#### Model 1

```
## chisq.scaled df.scaled cfi.scaled tli.scaled rmsea.scaled srmr
## 2930.988 804.000 0.899 0.892 0.062 0.063
```

## For constructs with categorical indicators, the alpha and the average variance extracted are calcula

```
## alpha 0.8344134 0.8650326 0.8954536 0.9077042 0.9057060 0.8694990  
## omega 0.8033959 0.8321705 0.8673721 0.8772246 0.8841962 0.8351232  
## omega2 0.8033959 0.8321705 0.8673721 0.8772246 0.8841962 0.8351232  
## omega3 0.8236923 0.8596939 0.8992562 0.9185328 0.9074923 0.8521143  
## avevar 0.4381763 0.4620095 0.5780030 0.5811473 0.5702036 0.6450147
```

#### Model 2 po usunięciu 3.3 i 4.3

```
## chisq.scaled df.scaled cfi.scaled tli.scaled rmsea.scaled srmr
## 2374.362 725.000 0.916 0.910 0.057 0.060
```

## For constructs with categorical indicators, the alpha and the average variance extracted are calcula

```
## alpha 0.8344134 0.8650326 0.8894462 0.9066283 0.9057060 0.8694990  
## omega 0.8022457 0.8307972 0.8564502 0.8720790 0.8839231 0.8355824  
## omega 0.8022457 0.8307972 0.8564502 0.8720790 0.8839231 0.8355824  
## omega 0.8202275 0.8546456 0.8825226 0.8984662 0.9069352 0.8521197  
## avevar 0.4376743 0.4619127 0.5986318 0.6067831 0.5706785 0.6448036
```

### Porównanie między modelami 1 i 2 MUSEQ

```
## [1] 475.027 0.000
```

### Model 2 po usunięciu 6.1

```
## chisq.scaled df.scaled cfi.scaled tli.scaled rmsea.scaled srmr
## 2185.913 687.000 0.922 0.916 0.056 0.058
```

## For constructs with categorical indicators, the alpha and the average variance extracted are calcula

```
## alpha 0.8344134 0.8650326 0.8894462 0.9066283 0.9057060 0.8493407

## omega 0.8027678 0.8310620 0.8565348 0.8722275 0.8840501 0.7784855

## omega2 0.8027678 0.8310620 0.8565348 0.8722275 0.8840501 0.7784855

## omega3 0.8219198 0.8556483 0.8827242 0.8989991 0.9071091 0.7935248

## avevar 0.4378164 0.4619508 0.5986320 0.6068100 0.5702232 0.6753243
```

## Porównanie między modelami 2 i 3 MUSEQ

**##** [1] 166.467 0.000

### Analiza BAPS

#### Model 1

##	chisq.scaled	df.scaled	cfi.scaled	tli.scaled	rmsea.scaled	srmr
##	1130.643	132.000	0.968	0.962	0.085	0.082

## For constructs with categorical indicators, the alpha and the average variance extracted are calcula

##		survival_strategy	negative_beliefs	normalizing_beliefs
##	alpha	0.9238154	0.9367321	0.9294762
##	omega	0.6783573	0.9137507	0.9129554
##	omega2	0.6783573	0.9137507	0.9129554
##	omega3	0.7118750	0.9250568	0.9475383
##	avevar	0.7049489	0.7278750	0.7247391

#### Model 2 bez BAPS1

```
## chisq.scaled df.scaled cfi.scaled tli.scaled rmsea.scaled srmr
## 903.224 116.000 0.974 0.970 0.081 0.070
```

## For constructs with categorical indicators, the alpha and the average variance extracted are calcula

##		<pre>survival_strategy</pre>	${\tt negative\_beliefs}$	normalizing_beliefs
##	alpha	0.9295837	0.9367321	0.9294762
##	omega	0.8858424	0.9137498	0.9130740
##	omega2	0.8858424	0.9137498	0.9130740
##	omega3	0.9048820	0.9250671	0.9477461
##	avevar	0.7414225	0.7279091	0.7247438

# Porównanie między modelami 1 i 2 BAPS

**##** [1] 200.528 0.000

## $\operatorname{MODEL}$ 3 z dodaną kowariancją między BAPS13 i BAPS14

##	chisq.scaled	df.scaled	cfi.scaled	tli.scaled r	cmsea.scaled	srmr
##	742.108	115.000	0.979	0.976	0.072	0.069

## For constructs with categorical indicators, the alpha and the average variance extracted are calcula

##		survival_strategy	${\tt negative\_beliefs}$	normalizing_beliefs
##	alpha	0.9295837	0.9367321	0.9294762
##	omega	0.8858410	0.9137363	0.8800942
##	omega2	0.8858410	0.9137363	0.8800942
##	omega3	0.9048735	0.9250082	0.8930653
##	avevar	0.7414095	0.7278819	0.6919380

# Porównanie między modelami 2 i 3 BAPS

**##** [1] 74.144 0.000