

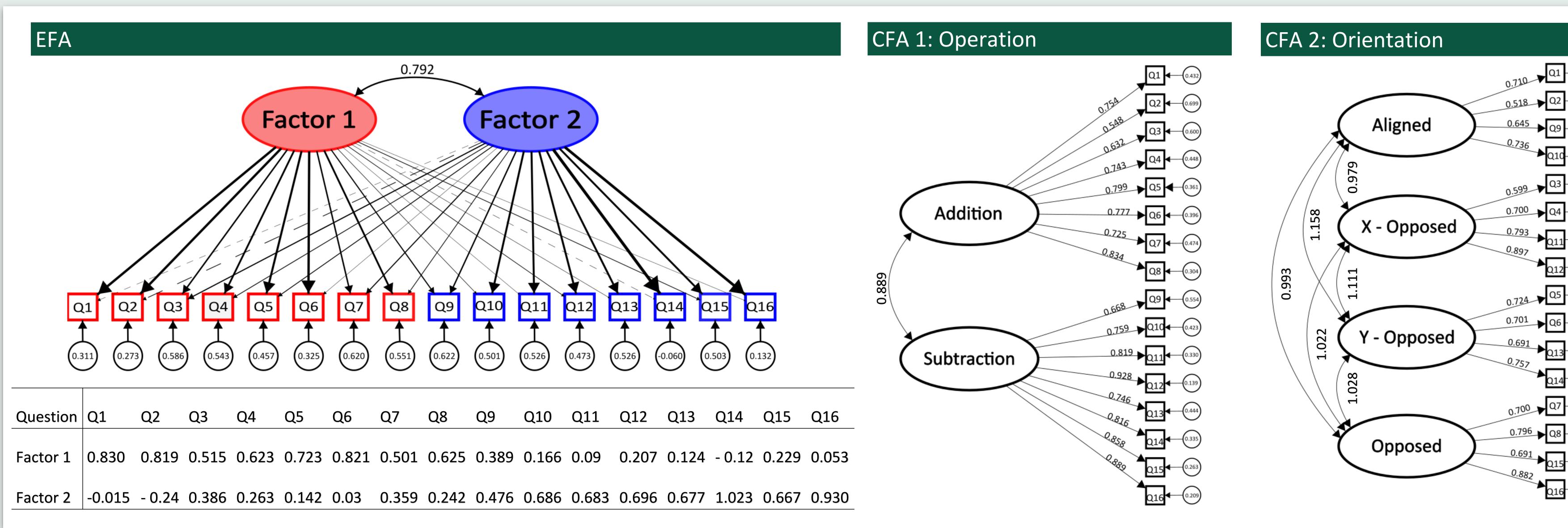
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#### Why Examine Vector Assessment

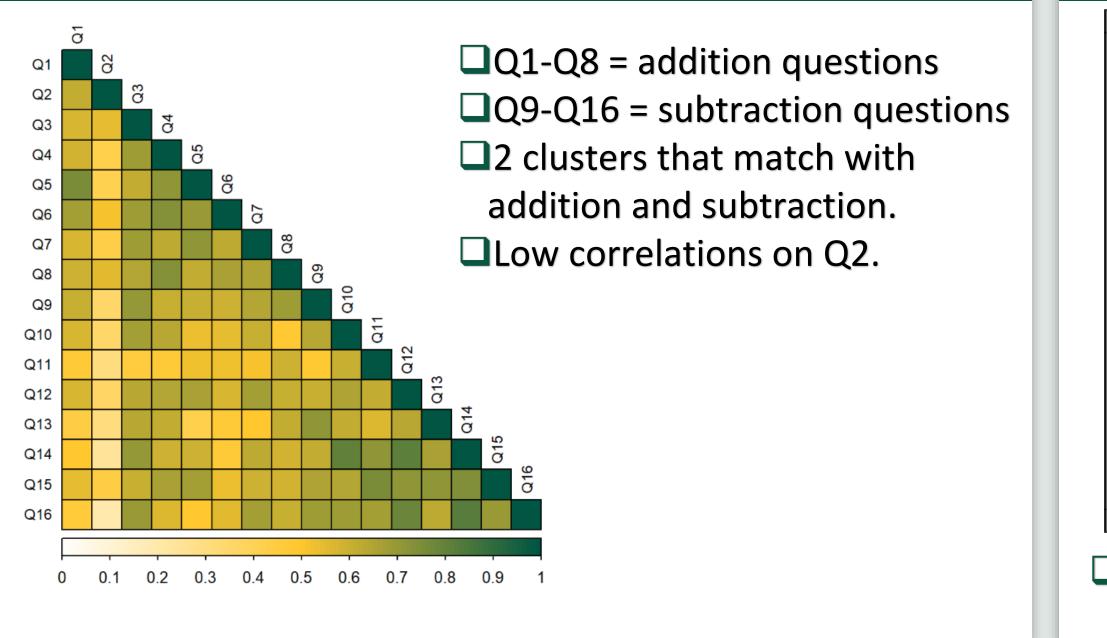
- Insufficient vector algebra skills are often linked to difficulties in introductory physics.
- Vector addition is well studied; yet there is little research on student abilities in vector subtraction.
- We are developing a tool to assess both vector addition and subtraction.

#### **Research Question:**

Can exploratory and confirmatory factor analysis (EFA and CFA) identify the factor structure and generate a model of what concepts our vector assessment is capturing?



#### **Correlation Matrix**

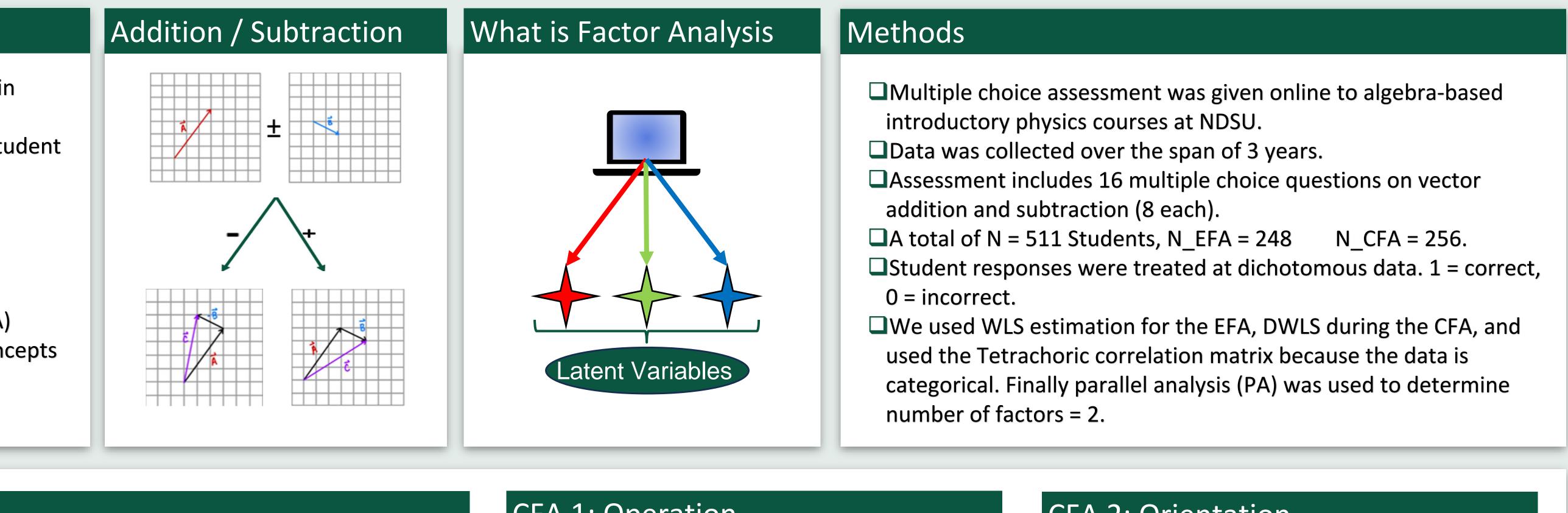


#### Acknowledgements

We would like to give a special thanks to Dan Bauer and Patrick Curran at Center Stat for supplying an intro course on SEMs. The course provided a conceptual understanding and groundwork for the methods used in the project. We would also like to thank the CiDER REU Cohort and Pollination Nation REU Cohort. Material based on work supported by NSF DUE 1852045. Any opinions, findings, conclusions, or recommendations expressed in this material are those of the authors and do not necessarily reflect the views of NSF.

# Using Factor Analysis to Explore the Structure of a Vector Assessment

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	Q10	Q11	Q12	Q13	Q14	Q15	Q16
89	0.166	0.09	0.207	0.124	- 0.12	0.229	0.053
76	0.686	0.683	0.696	0.677	1.023	0.667	0.930

quation of the Model						
$\begin{bmatrix} Q 1 \end{bmatrix}$		0.754	0		0.432	
Q 2		0.548	0		0.699	
Q 3		0.632	0		0.600	
<i>Q</i> 4		0.743	0	[ Addition [ Subtraction ] +	0.448	
Q 5		0.799	0		0.361	
Q 6		0.777	0		0.396	
Q 7		0.725	0		0.474	
Q8	_	0.834	0		0.304	
Q 9	=	0	0.668		0.554	
<i>Q</i> 10		0	0.759		0.423	
Q 11		0	0.819		0.330	
<i>Q</i> 12		0	0.928		0.139	
<i>Q</i> 13		0	0.746		0.444	
<i>Q</i> 14		0	0.816		0.335	
Q 15		0	0.858		0.264	
<i>Q</i> 16		0	0.889		0.209	
Intercept is set to zero in model						
identification.						

### 

Results	& Discuss	sion						
Ctatiatia	·/2 / DF	<b>T</b> 11				Coverience	Eigenvalues	CFA 2
Statistic	χ <sup>2</sup> / DF	TLI	RMSEA	CFI	SRMR	Covariance	Factor(1:2)	AL
EFA	35.51 >> 2	0.692 << 0.9	0.207 > 0.2			0.792	5.76 : 4.98	Х - Орр
CFA 1	1.11 < 2	0.998 >> 0.9	0.021 << 0.2	0.998 >> 0.9	0.071	0.881	0.71:0.12	Ү — Орр
CFA 2	1.31 < 2	0.995 >> 0.9	0.035 << 0.2	0.996 >> 0.9	0.075	See above	See right	Орр
🗖 PA sua	gests 2 facto	ors.						

□ EFA results suggest an +/- split. EFA statistical values are not significant.

<sup>2nd</sup> CFA is not a possible model.

All statistical values are significant; yet covariance matrix in not positive definite. Additionally, the covariances between all factors in the model are extremely high.

□ The 1<sup>st</sup> CFA is a possible factor model. The statistics of the 1<sup>st</sup> CFA are all significant. • Our assessment tests two separate topics as suspected. More importantly these two topics are vector addition

and subtraction. fewer parameters would produce a closer fitting model. We hope to investigate this concern more to validate the

U We would also like to report our potential concerns with the first CFA model. It is unexpected that the CFA with conclusions of the study.

U We are not confident in the extraction of eigenvalues from the CFA models. We plan to investigate this further.



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0.495
0.732
0.584
0.458
0.641
0.510
0.371
0.196
0.476
0.509
0.523
0.427
0.510
0.366
0.278
0.221

