



Computational Thinking

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Briefing Agenda

- Background and Project Overview
 - Phase 1: Computational Thinking Score
 - Phase 2: Computational Thinking Score Validation
- Predictors and Criterion
- Phase 2 Results
- Closing

Congressional Mandate

- William M. (Mac) Thornberry National Defense Authorization Act (NDAA) for Fiscal Year 2021 (HR 6395), Section 594
 - Must assess six (6) computational thinking construct domains
 - Problem Decomposition
 - Abstraction
 - Pattern Recognition
 - Analytical Ability
 - Identifying Variables for Data Representation
 - Creating Algorithms and Solution Expressions
 - Must be available for operational use by October 1, 2024

Computational Thinking Construct Domains

Construct Domains	Descriptions
1. Problem decomposition	▪ Break down a problem/task into smaller/easier components (e.g., describe a system as a sequence of processes)
2. Abstraction	▪ Focus on the most relevant information and ignore extraneous information to interpret meaning and reduce complexity of a problem/task
3. Pattern recognition	▪ Identify and use repeated information or patterns to predict outcomes or determine actions for a problem/task
4. Analytical ability	▪ Inspect, cleanse, transform, and model data with the goal of discovering useful information for a problem/task
5. Identifying variables for data representation	▪ Recognize how parts of a solution may be reapplied to, or eliminated from, similar or unique problems/tasks
6. Creating algorithms and solution expressions	▪ Recognize and evaluate options against outcomes to simplify or automate processes for efficiency and resource utilization improvements

Where We Started

- Existing measures of computational thinking were not viable
 - Those used for selection require specific programming language skills
 - Those used for skill acquisition are developed for the K–12 classroom environment, which are free on the internet (lack test security)
- NDAA-specified deadline of 01 October 2024 did not support creating a new, valid measure of computational thinking
- Belief that the Complex Reasoning Test (CR) already under development, and possibly some of the ASVAB subtests [e.g., Arithmetic Reasoning (AR), Assembling Objects (AO)] and other special tests [e.g., Cyber (CT), Coding Speed (CS), Mental Counters (MCt)] were likely assessing the computational thinking construct domains

Project Overview

Phase 1: Define Computational Thinking Score Equation

- Gather empirical & SME-estimated correlations
- Specify & analyze prediction models
- Generate, evaluate, finalize synthetic CompT score equations
- Submit software requirements & specifications

Phase 2: Verify Validity of Computational Thinking Scores

- Select computational thinking marker test
- Develop & implement data collection plan at MEPS
- Match shippers' ASVAB & CT scores to study data & clean
- Conduct analyses & summarize results

Computational Thinking Score Equations

CompT_AR

$$2CR + AR$$

CompT_CT

$$2CR + CT$$

CompT_ALL

$$2CR + AR + CT$$

Note: Scores are a weighted sum of CR, AR, and CT standard (T) scores with $X = 50$, $std = 10$. The AR, CR, and CT standard (T) scores are normed to the PAY97 sample.

Validation Data Collection

Collected Data	Matched Data	Cleaned Data
<ul style="list-style-type: none">■ MEPS administered the Qualtrics data collection tool between 4/15 – 5/20<ul style="list-style-type: none">✓ Complex Reasoning (CR)✓ Computational Thinking Assessment for Middle Schoolers (CTA-M)✓ Background questions■ Shippers = 1,044	<ul style="list-style-type: none">■ HumRRO sent DTAC participant IDs from Qualtrics on weekly basis■ DTAC used participant IDs and MEPS rosters to pull ASVAB and CT scores into a de-identified dataset■ HumRRO appended with responses on CR, CTA-M, and background questions■ Shippers = 922	<ul style="list-style-type: none">■ Removed any that showed a lack of motivation using:<ul style="list-style-type: none">✓ Two CR attention-check items✓ Self-report question at end✓ Time spent on CR and CTA-M (no more than 2 standard deviations below the mean for time spent)✓ Checks for careless response patterns✓ Checks for CR and CTA-M scores that were at or below chance■ Removed any that left study early for transportation■ Shippers = 722

Sample by Demographic Group

Gender		Race-Ethnicity		Service*	
Female	106	Hispanic White (HW)	166	Air Force	232
Male	608	Non-Hispanic Asian (NHA)	35	Army	22
NA	8	Non-Hispanic Black (NHB)	172	Coast Guard	0
		Non-Hispanic White (NHW)	291	Marine Corps	214
		Other or NA	58	Navy	238
				Space Force	16
Total	722		722		722

*Participation was limited to Shippers with a pre-enlistment CT score. Therefore, an equal distribution across Services was not expected given Services have different policies for administering CT to applicants.

Sample by Type of Service and Component

Component	Service						Total
	Army	Air Force	Coast Guard	Marine Corps	Navy	Space Force	
Active Duty	21	232	0	205	235	16	709
Guard	1	0	0	9	3	0	13
Reserve	0	0	0	0	0	0	0
Total	22	232	0	214	238	16	722

Predictors and Criterion

Overview of Predictors

- Components of operational equation-based Computational Thinking scores
 - AR
 - CT
 - CR
- Operational equation-based Computational Thinking scores derived from Phase 1 study
 - $\text{CompT_AR} = 2\text{CR} + \text{AR}$
 - $\text{CompT_CT} = 2\text{CR} + \text{CT}$
 - $\text{CompT_ALL} = 2\text{CR} + \text{AR} + \text{CT}$

Overview of Criterion

- Computational Thinking Abilities – Middle Grades Assessment (CTA-M)
 - Developed by Wiebe et al, 2019
 - Designed for classroom use with middle school students
- Consists of 23 items administered with a 45-minute time limit
 - 15 Computational Thinking Test (CTt) items (Gonzalez et al., 2015)
 - 8 Bebras items (2016 UK Bebras Challenge)
- Items map to two or three of the six construct domains based on consensus judgments by HumRRO team members
 - Problem Decomposition
 - Solving for Algorithms
 - Analytical Ability

Predictor and Criterion Analyses

- Calculated score for each Shipper on:
 - CTA-M (criterion)
 - CR (predictor)
- Calculated the three CompT scores using the operational equations from Phase 1
 - CompT_AR
 - CompT_CT
 - CompT_ALL
- Computed predictor and criterion descriptive statistics
- Computed predictor and criterion reliability estimates (except AR and CT*)
- Computed predictor and criterion subgroup differences (except AR and CT*)

*For AR and CT, used existing estimates of reliability documented in psychometric checklists (Sinclair et al., 2003) and current estimates of subgroup differences for FY23 applicant data (Johnston-Fisher et al., 2024).

Predictor and Criterion Descriptives

Variable	Variable Type	Mean	Median	SD	Min	Max
CTA-M	Criterion	13.8	14.0	4.1	6	23
AR	Predictor	52.8	52.5	7.9	30	72
CT	Predictor	51.6	52.0	8.9	22	76
CR	Predictor	55.0	57.0	8.1	35	67
CompT_AR	Equation Score	162.8	167.0	20.8	103	202
CompT_CT	Equation Score	161.5	164.0	20.5	104	202
CompT_ALL	Equation Score	214.3	217.0	25.6	141	271

Predictor and Criterion Reliabilities

Variable	Type of Variable	Reliability	
		Cronbach's Alpha	Mosier's Composite Formula
CTA-M	Criterion	0.73	—
AR*	Predictor	0.89	—
CT*	Predictor	0.70	—
CR	Predictor	0.82	—
CompT_AR	Equation Score	—	0.88
CompT_CT	Equation Score	—	0.83
CompT_ALL	Equation Score	—	0.88

*Cronbach's alpha obtained from Psychometrics Checklists as reported in Sinclair et al. (2023).

Predictor and Criterion Subgroup Differences

Variable	Type of Variable	Cohen's <i>d</i> Effect Size			
		M-F	NHW-NHB	NHW-HW	NHW-NHA*
CTA-M	Criterion	0.28	0.54	0.19	0.35
AR	Predictor	0.25	0.43	0.10	-0.01
CT	Predictor	0.44	0.36	0.21	0.18
CR	Predictor	0.07	0.23	0.11	-0.01
CompT_AR	Equation Score	0.15	0.34	0.12	-0.01
CompT_CT	Equation Score	0.25	0.34	0.18	0.08
CompT_ALL	Equation Score	0.28	0.41	0.17	0.06

Effect Size Category	
Less than Small	<0.20
Small	0.20 – 0.49
Moderate	0.50 – 0.79

*Sample size for Non-Hispanic Asian subgroup is too small to support interpretation of effect sizes.

Phase 2 Results

Data Analysis Plan

- Calculate zero-order correlations between CTA-M and the three components (AR, CT, CR) in the three Computational Thinking score equations
 - Correct results for range restriction
 - Disattenuate results for criterion unreliability
- Calculate zero-order correlations between CTA-M and the three operational equation-based Computational Thinking scores developed in Phase 1
 - Correct results for range restriction
 - Disattenuate results for criterion unreliability
- Estimate empirical validity of non-negative least square (NNLS) regression equations using data from Phase 2 validation study
 - Correct results for range restriction
 - Disattenuate results for criterion unreliability
 - Adjust results for shrinkage
- Conduct post-hoc analysis to recompute estimates using all 9 ASVAB subtests, CT, and CR

Correlation of Equation Component Tests with CTA-M

Equation Component Test	Correlation with CTA-M	
	Observed	Corrected*
AR	0.48	0.71
CT	0.40	0.61
CR	0.54	0.73

*Results are corrected for multivariate range restriction and disattenuated for criterion unreliability.

Correlation of Operational Equation-based Scores with CTA-M

Operational Equation-Based Score	Correlation with CTA-M	
	Observed	Corrected*
CompT_AR	0.61	0.81
CompT_CT	0.60	0.80
CompT_ALL	0.63	0.83

*Results are corrected for multivariate range restriction and disattenuated for criterion unreliability.

NNLS Regression Results by Operational Equation Scores

Regression Coefficient/ Multiple R	CompT_AR		CompT_CT		CompT_ALL	
	Observed	Corrected*	Observed	Corrected*	Observed	Corrected*
AR	0.15	0.18	—	—	0.11	0.13
CT	—	—	0.12	0.15	0.09	0.09
CR	0.2	0.23	0.23	0.27	0.20	0.22
Multiple R	0.61	0.81	0.60	0.80	0.63	0.83
R Shrinkage	0.61	0.81	0.60	0.80	0.63	0.83

*Results are corrected for multivariate range restriction and disattenuated for criterion unreliability.

Operational vs. NNLS Regression Validity Results

Computational Thinking Score	Validity Estimates			
	Operational Equations Based on Phase 1 Synthetic Validity Study		NNLS Regression Equations Based on Phase 2 Criterion-Related Validity Study	
	Observed	Corrected*	Observed	Corrected*
CompT_AR	0.61	0.81	0.61	0.81
CompT_CT	0.60	0.80	0.60	0.80
CompT_ALL	0.63	0.83	0.63	0.83

*Results are corrected for multivariate range restriction and disattenuated for criterion unreliability.

Post-Hoc Validity Estimates with All ASVAB Subtests + CT + CR

Computational Thinking Score	Validity Estimates			
	Operational Equations Based on Phase 1 Synthetic Validity Study		NNLS Regression Equations Based on Phase 2 Criterion-Related Validity Study	
	Observed	Corrected*	Observed	Corrected*
CompT_AR	0.61	0.81	0.61	0.81
CompT_CT	0.60	0.80	0.60	0.80
CompT_ALL	0.63	0.83	0.63	0.83
Post-hoc = All ASVAB subtests + CT + CR	--	--	0.67	0.87

*Results are corrected for multivariate range restriction and disattenuated for criterion unreliability. Yellow highlights identify post-hoc results to use for comparison to empirical results for computational thinking scores (same as slide 23).

Results Conclusion

- All three equation-based scores (CompT_AR, CompT_CT, CompT_ALL) were strong predictors of the computation thinking construct, at least as it was operationalized in the Phase 2 validity study (i.e., CTA-M)
- Empirical weights for the score components (AR, CT, CR) derived from the Phase 2 validity study did not outperform the operational weights derived from the Phase 1 synthetic validity study
- Empirical validity estimates using all ASVAB subtests, CT, and CR resulted in relatively small increases ($\Delta R = 0.04$) in prediction in CTA-M scores

Closing

Software Updates (Completed)

- CR is available for administration on the iCAT platform
- Applicant's completion of CR triggers calculation of CompT scores
 - Requires an AR and/or CT score within the last 2 years
 - Uses most recent AR and/or CT score when multiple records are found
 - Submits a blank score if an eligible AR and/or CT score is not found
- Saves each CompT score within the applicant's CR record
- MEPCOM receives all 4 scores: CR as well as 3 CompT scores

Questions?



Response to June 2024 DAC Recommendation

- In process of preparing research designs for CR and CompT that DTAC may consider for future research
 - Applicant data containing one to three of the CompT scores is slowly accumulating, which will support additional analyses
 - Demographic information will likely be available for future subgroup differences research
 - Shippers' occupational training criteria may be useful for future research, should it be made available
 - ASVAB Training Relevance Survey results may be used to identify military occupations with high computational thinking relevance results to further research

Questions to DAC

- Does the DAC have any suggestions for conducting additional research on fairness issues and/or validity?

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Thank you!

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