



Background

Children's drawings have a long and diverse history of use as primarily nonverbal assessment of children's cognitive, emotional, and motor function. The most common approach to assign progress or delay in children's development when using drawing tasks are checklists marking the presence or absence of intuitively-derived features. The most influential drawing checklist is described by Florence Goodenough in her 1926 book, Measurement of intelligence by drawings which outlines the Draw-A-Man test, a checklist system to identify important features of human figures (e.g., body parts, facial features).

More recent research on drawings has extended the checklist approach to assess shape drawings (e.g., cubes). In the present research, we explore how recent advances leveraging neural network models and crowd-sourced perceptual judgements may be applied to cube drawings. We consider the predictive relationship of children's cube and human figure drawings on the cognitive/motor task of paper folding.













Does the checklist approach underrepresent structure in children's drawings?



(Mitchelmore, 1978)



(Cox & Perara, 1998)

	\bigcirc
\square	Ē

(Lange-Küttner & Ebersbach, 2013)

<u>Score 8</u> – Nice 3D Cube

The Current Study



129 Children from age 2 to 12, 53% female; 25 Adults 65 Children contributed a Cube Drawing / Paper Folding $M_{age} = 6$ years; Range_{age} = 4 – 12 years

- 65% female
- **Recruited for a larger study looking at drawing on different media** (Kirkorian et al., 2020)

Unfolding Structure in the Drawings of Cubes Clint A. Jensen¹, Timothy T. Rogers¹, Brittany G. Travers¹, Heather L. Kirkorian¹ & Karl S. Rosengren² ¹University of Wisconsin-Madison, ²University of Rochester

Contemporary approaches (Jensen et al., 2023; 2024)





Note. Each drawing is placed according to its coordinates in the corresponding 2D space, with the color indicating the Quality Rank (QR) score of the drawing, with hotter colors indicating higher-ranked drawings and cool colors showing low-ranked drawings.

Correlations of Checklist Scores and Drawing Quality







Dependent Variable	n	Baseline	Cube Checklist	Cube Quality Ranking	Cube Human Embedding	Cube Machine Embedding	Cube QR * Human Embedding	Cube QR * Machine Embedding
Offset	41	<i>adj</i> = .06 <i>mult</i> = .09	<i>adj</i> = .18 <i>mult</i> = .20	<i>adj</i> = .10 <i>mult</i> = .26	adj = .26* mult = .43*	<i>adj</i> = .17 <i>mult</i> = .56	<i>adj</i> = .41 <i>mult</i> = .78	<i>adj</i> = .76 <i>mult</i> = .99
Angle	41	<i>adj</i> = .00 <i>mult</i> = .08	adj = .11* mult = .13*	adj = .20* mult = .34*	adj = .16 mult = .47	<i>adj</i> = .14 <i>mult</i> = .48	adj = .48* mult = .72*	adj = .41 mult = .97
comparison model: null		Baseline			Quality Ranking *Baseline			

Human Figure Drawings

Dependent Variable	п	Baseline	Human Figure Checklist	Human Figure Quality Ranking	Human Figure Human Embedding	Human Figure Machine Embedding	Human Figure QR * Human Embedding	Human Figure QR * Machine Embedding
Offset	34	adj = .10* mult = .13*	adj = .28 mult = .45	adj = .31 mult = .47	adj = .51* mult = .72*	adj = .64* mult = .87*	adj = .63* mult = .82*	adj = .56 mult = .78
Angle	34	<i>adj</i> = .10 <i>mult</i> = .19	adj = .51* mult = .63*	adj = .38* mult = .53*	<i>adj</i> = .49 <i>mult</i> = .68	<i>adj</i> = .28 <i>mult</i> = .73	adj = .42 mult = .77	adj = .67* mult = .83*
comparison model: null		Baseline			Quality Ranking *Baseline			

Note. All models include age and gender and their interactions as covariates of no interest. Baseline models additionally include the Cube Quality Ranking (QR) score. All models include all interactions. Asterisks indicate statistical significance at '*' p < 0.05.



References

Psychology, 18(3), 309-317.

Jensen, C. A., Sumanthiran, D., Kirkorian, H. L., Travers, B. G., Rosengren, K. S., & Rogers, T. T. (2023). Human perception and machine vision reveal rich latent structure in human figure drawings. Frontiers in Psychology, 14, 1029808. doi/10.3389/fpsyg.2023.1029808

Jensen, C. A., Rogers, T. T., & Rosengren, K. S. (2024). Better than Goodenough? Evaluating new computational techniques for finding diagnostic structure in children's drawings. *Memory & Cognition*, 1-19. doi/10.3758/s13421-024-01557-0

Kirkorian, H. L., Travers, B. G., Jiang, M. J., Choi, K., Rosengren, K. S., Pavalko, P., & Tolkin, E. (2020). Drawing across media: A crosssectional experiment on preschoolers' drawings produced using traditional versus electronic mediums. Developmental Psychology, 56(1), 28. doi/10.1037/dev0000825

Lange-Küttner, C., & Ebersbach, M. (2013). Girls in detail, boys in shape: Gender differences when drawing cubes in depth. British Journal of Psychology, 104(3), 413-437.

Mitchelmore, M. C. (1978). Developmental stages in children's representation of regular solid figures. *The Journal of Genetic* Psychology, 133(2), 229-239.



$offset = \frac{|C1| + |C2|}{2}$

angle = |C2 - C1|

Used stepwise regression including our novel metrics to predict paper folding scores

Results varied by media

Marker on Paper > Stylus on Tablet >> Finger on Tablet

Findings suggest an association between the fine motor U skills used in paper folding and the dexterity required to hold and control a marker/pen while drawing.

Contemporary methods increase ability to predict latent structure in children's drawings.

Cox, M. V., & Perara, J. (1998). Children's Observational Drawings: a nine-point scale for scoring drawings of a cube. *Educational*