

Dependency learning: comparing adults and children

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Comparisons between children and adults

In this final analysis we evaluate differences between the adults (Experiment 1) and the children results (Experiment 2). In particular, we focused on the effect of main interest to evaluate implicit learning; i.e. the slowdown in reaction times in the transfer block (Block 7) relative to Block 6. To compared the data from adults and children we pooled the serial reaction-time data from both experiments and extracted the data from Block 6 and 7.

This analysis was largely similar to the reaction-time models reported earlier: reaction times were modelled in Bayesian linear mixed-effects models. Reaction times for elements that were not part of an adjacent or nonadjacent dependency were included in the model as baseline. Model predictors were Age Group (levels: adults, children), Dependency (levels: dependency, baseline), Adjacency (levels: adjacent, nonadjacent), Block (levels: 6, 7), and all by-Dependency and by-Adjacency interactions with Group and Block. All predictors were sum-coded following Schad et al. (2020). Reaction time data were fitted with a shifted-lognormal distribution, random intercepts for each stimulus image, and participants with by-participant slope adjustments for all main effects and interactions.

In addition to the parameter estimates as in the previous results sections, we used the posterior to calculate the standardised effect sizes δ , defined as $\delta = \frac{\mu}{\sigma}$, where μ is the parameter value of the effect of interest and σ is the variance. The effect size was calculated to allow comparisons across age group (Wagenmakers et al., 2010). Because the variance is necessary to calculate the effect size, we specified the model with two different variance components, one for the children group and one for the adults group as the population standard deviation is likely to be different for each group (Baguley, 2009).

For the effect sizes, we determined the region of practical equivalence (henceforth, ROPE) to assess the uncertainty of the effect size (Makowski et al., 2019). The ROPE is the

range of values that are practically equivalent to a null effect (see e.g. Kruschke, 2010, 2011). For the effect size we set the ROPE to be -0.1 and 0.1 (Kruschke, 2018) which is the range of negligible effect sizes (Cohn, 1988). The value returned is the proportion of posterior samples (of the effect size) that falls inside the ROPE. In other words, the ROPE value indicates the extent to which the posterior cannot rule out an negligible effect. A meaningful effect size should have a small proportion of posterior samples within the ROPE.

Table 1 summaries the modelling outcome for the serial reaction-time data. We found compelling evidence for longer reaction times for children compared to adults. Reaction times for Block 7 were longer than for Block 6. Further evidence was found for a two-way interaction of Age Group and Dependency. Importantly, we found evidence for a three-way interaction of Dependency, Block and Age Group. The variance estimate for the children group ($\hat{\sigma} = 0.37$, 95% HPDI[0.36, 0.38]) was larger than the variance estimate for the adults group ($\hat{\sigma} = 0.28$, 95% HPDI[0.27, 0.29]).

Figure 1 illustrates the modelled serial reaction-time data comparing the slowdown from Block 6 to Block 7 for the adult and the children group. The slowdown from Block 6 to 7 was tested within Age Group and Dependency / Adjacency to gain further insight into the source of the two three-way interactions. Reported are the effect sizes for each comparison.

We observed an interaction of Dependency (levels: dependency, baseline), Block, and Age Group. For the adult group, we observed a small effect for a slowdown from Block 6 to 7 in the dependency condition ($\hat{\delta} = 0.27$, 95% HPDI[0.11, 0.43], ROPE = 0%) but negligible evidence for a slowdown in baseline trials ($\hat{\delta} = -0.01$, 95% HPDI[-0.11, 0.09], ROPE = 100%). For the children group, we observed negligible to small slowdown effects in both the dependency condition ($\hat{\delta} = 0.14$, 95% HPDI[-0.01, 0.28], ROPE = 35%) and for the baseline condition ($\hat{\delta} = 0.14$, 95% HPDI[0.04, 0.25], ROPE = 16%).

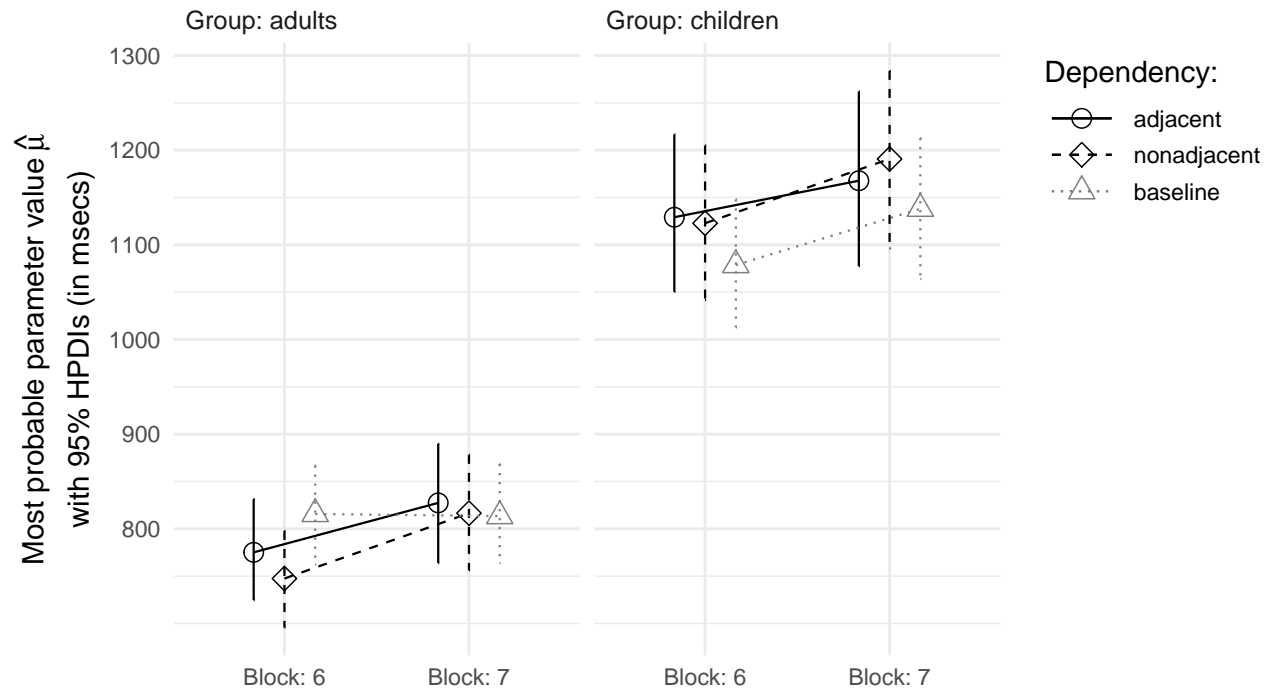


Figure 1

Modelled serial reaction-time data (group comparison). Reaction-time data are shown for each age group for the final two blocks (block 7 is the transfer block). Shown are the modelled reaction-time data with 95% HPDI (in msecs) for both adjacency conditions (adjacent, nonadjacent) and the baseline condition.

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Table 1

Age-group comparison for serial reaction-time data.

Estimated parameter values (in msec) for main effects and interactions of Age Group (levels: adults, children), Dependency (levels: dependency, baseline), Adjacency (levels: adjacent, nonadjacent).

| Predictor | $\hat{\mu}$ | 95% HPDI | $P(\hat{\mu} < 0)$ |
|--|-------------|--------------|--------------------|
| Main effects | | | |
| Age Group | -390 | [-458, -330] | >.999 |
| Dependency | 11 | [-30, 47] | .322 |
| Adjacency | -8 | [-37, 21] | .703 |
| Block | -67 | [-104, -34] | >.999 |
| Two-way interactions | | | |
| Dependency \times Block | -33 | [-74, 11] | .934 |
| Adjacency \times Block | -12 | [-42, 19] | .769 |
| Age \times Block | -1 | [-34, 32] | .52 |
| Age \times Dependency | -65 | [-94, -31] | >.999 |
| Age \times Adjacency | -14 | [-43, 11] | .856 |
| Three-way interactions | | | |
| Dependency \times Block \times Age | -39 | [-73, -6] | .99 |
| Adjacency \times Block \times Age | 2 | [-28, 29] | .479 |

Note. $\hat{\mu}$ = most probable parameter value; 95% HPDI = interval containing 95% of the probability mass;
 $P(\hat{\mu} < 0)$ = probability of the true parameter value being smaller than 0.