Not All Procedural Learning Tasks Are Difficult for Adults With Developmental Language Disorder

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Purpose: The experiment reported here compared two hypotheses for the poor statistical and artificial grammar learning often seen in children and adults with developmental language disorder (DLD; also known as specific language impairment). The procedural learning deficit hypothesis states that implicit learning of rule-based input is impaired, whereas the sequential pattern learning deficit hypothesis states that poor performance is only seen when learners must implicitly compute sequential dependencies. The current experiment tested learning of an artificial grammar that could be learned via feature activation, as observed in an associatively organized lexicon, without computing sequential dependencies and should therefore be learnable on the sequential pattern learning deficit hypothesis, but not on the procedural learning deficit hypothesis.

Method: Adults with DLD and adults with typical language development (TD) listened to consonant–vowel–consonant–vowel familiarization words from one of two artificial phonological grammars: Family Resemblance (two out of three features) and a control (exclusive OR, in which both consonants are voiced OR both consonants are voiceless) grammar in which no learning was predicted for either group. At test, all participants rated 32 test words as to whether or not they conformed to the pattern in the familiarization words.

Results: Adults with DLD and adults with TD showed equal and robust learning of the Family Resemblance grammar, accepting significantly more conforming than nonconforming test items. Both groups who were familiarized with the Family Resemblance grammar also outperformed those who were familiarized with the OR grammar, which, as predicted, was learned by neither group.

Conclusion: Although adults and children with DLD often underperform, compared to their peers with TD, on statistical and artificial grammar learning tasks, poor performance appears to be tied to the implicit computation of sequential dependencies, as predicted by the sequential pattern learning deficit hypothesis.

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The research presented here employs artificial grammar learning to contrast the procedural learning and sequential pattern learning deficit accounts. There is a growing body of evidence that children and adults with DLD often perform relatively poorly in both statistical learning and artificial grammar learning experiments (e.g., Evans et al., 2009; Grunow et al., 2006; Lukács & Kemény, 2014; Obeid et al., 2016; Plante et al., 2002). For example, in a statistical learning experiment, learners are familiarized with strings of nonwords such as “dutaba, tutibu, pidabu, patubi, bupada, and babupu” with no silence between words (Saffran et al., 1996). At test, they must guess which new items were “words” during familiarization, a task thought to be accomplished through the use of high (often 100%) sequential dependencies between adjacent syllables in the familiarization words. Children with DLD perform more poorly on these tasks than do their peers with TD (e.g., Evans et al., 2009; Obeid et al., 2016). A similar requirement for detecting dependency relations can be seen in many artificial grammar learning experiments. In one such experiment, learners were exposed to strings of the form aXb or cXd, in which the syllable that occurs in the third position (from either a b-set or a d-set of syllables) is dependent on what occurs in the first position (from either an a-set or a c-set of syllables, respectively; Gómez & Maye, 2005). Adult college students with DLD showed poorer learning of this language than did adults with TD (Grunow et al., 2006).

Note that all of these studies in which children or adults with DLD perform poorly would be considered procedural learning of rule-governed input by the procedural learning deficit hypothesis and sequential dependency learning tasks by the sequential pattern learning deficit hypothesis. How might we differentiate these two accounts? As noted above, the procedural learning deficit hypothesis predicts poor performance by participants with DLD for any learning study in which performance is dependent on implicitly detecting and using the underlying rule- or principle-governed structure of the input. In contrast, the sequential pattern learning deficit hypothesis only predicts poorer performance by participants with DLD if the underlying structure requires detecting sequential dependencies in particular. Therefore, the latter hypothesis predicts TD-level performance in artificial grammar learning experiments that do not require detecting sequential dependencies.

Published experiments that involve the implicit learning of stress assignment rules of an artificial grammar appear to support the sequential pattern learning deficit hypothesis over the procedural learning deficit hypothesis. In two such experiments, participants heard a set of familiarization words generating experiment, learners are familiarized with strings of nonwords such as “dutaba, tutibu, pidabu, patubi, bupada, and babupu” with no silence between words (Saffran et al., 1996). At test, they must guess which new items were “words” during familiarization, a task thought to be accomplished through the use of high (often 100%) sequential dependencies between adjacent syllables in the familiarization words. Children with DLD perform more poorly on these tasks than do their peers with TD (e.g., Evans et al., 2009; Obeid et al., 2016). A similar requirement for detecting dependency relations can be seen in many artificial grammar learning experiments. In one such experiment, learners were exposed to strings of the form aXb or cXd, in which the syllable that occurs in the third position (from either a b-set or a d-set of syllables) is dependent on what occurs in the first position (from either an a-set or a c-set of syllables, respectively; Gómez & Maye, 2005). Adult college students with DLD showed poorer learning of this language than did adults with TD (Grunow et al., 2006).

Published experiments that involve the implicit learning of stress assignment rules of an artificial grammar appear to support the sequential pattern learning deficit hypothesis over the procedural learning deficit hypothesis. In two such experiments, participants heard a set of familiarization words created based on principles such as “stress heavy syllables” and “stress final syllables.” The principles were ranked with respect to each other, such that, if two principles could be applied to the same input yielding different outcomes, the most important (highly ranked, e.g., “stress heavy syllables”) principle is applied. Importantly, learning the principles did not entail detecting sequential dependencies among parts of the word (i.e., none of the rules had dependencies such as “stress the first syllable if the last syllable is heavy”). Children and adults with DLD showed significant learning and did not perform differently from peers with TD (Bahl et al., 2009; Plante et al., 2010). These results are consistent with the sequential pattern learning deficit hypothesis, because there were no sequential dependencies in the stimuli. The current work. The procedural learning deficit hypothesis adopts the view that there are at least two distinct memory systems, namely, a procedural system that is slow and implicit and a declarative system that is fast and explicit. Ullman and Pierpont (2005, p. 403) say of the two systems with respect to language: “According to this view—referred to as the Declarative/Procedural (DP) model—idiosyncratic mappings are stored in a memorized ‘mental lexicon’ that depends on declarative memory, whereas the learning and use of rule-governed computations involves a ‘mental grammar’ that depends on procedural memory.” We have elsewhere (Goffman & Gerken, 2020; Plante, 2020) outlined problems with this particular pairing of dichotomies (fast–explicit, slow–implicit) as they apply to rapid rule learning in both human infants (e.g., Gerken & Knight, 2015) and some nonhuman animals (e.g., Smith et al., 2012). For present purposes, however, the important point is that, on the procedural learning deficit account, any rule- or principle-governed input that is learned implicitly and that cannot be memorized or stored as chunks should be affected in DLD (e.g., Ullman & Pierpont, 2005). In contrast, the sequential pattern learning deficit hypothesis does not adopt the procedural–declarative dichotomy but rather focuses specifically on the learning of sequential dependencies. In general, these dependencies constitute a set that overlaps with the set that falls under the procedural learning umbrella, since not all rule- or principle-governed input entails sequential dependencies and not all sequential input is rule governed (see Footnote 1).

The research presented here employs artificial grammar learning to contrast the procedural learning and sequential pattern learning deficit accounts. There is a growing body of evidence that children and adults with DLD often perform relatively poorly in both statistical learning and artificial grammar learning experiments (e.g., Evans et al., 2009; Grunow et al., 2006; Lukács & Kemény, 2014; Obeid et al., 2016; Plante et al., 2002). For example, in a statistical learning experiment, learners are familiarized with strings of nonwords such as “dutaba, tutibu, pidabu, patubi, bupada, and babupu” with no silence between words (Saffran et al., 1996). At test, they must guess which new items were “words” during familiarization, a task thought to be accomplished through the use of high (often 100%) sequential dependencies between adjacent syllables in the familiarization words. Children with DLD perform more poorly on these tasks than do their peers with TD (e.g., Evans et al., 2009; Obeid et al., 2016). A similar requirement for detecting dependency relations can be seen in many artificial grammar learning experiments. In one such experiment, learners were exposed to strings of the form aXb or cXd, in which the syllable that occurs in the third position (from either a b-set or a d-set of syllables) is dependent on what occurs in the first position (from either an a-set or a c-set of syllables, respectively; Gómez & Maye, 2005). Adult college students with DLD showed poorer learning of this language than did adults with TD (Grunow et al., 2006).

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1Here, we use the term statistical learning to refer to those experiments in which the input stimuli are not generated by rules or principles and that require learners to use transitional probabilities for word segmentation. We use the term artificial grammar learning to refer to experiments in which the stimuli are generated by rules or principles and in which learners are tested for having learned those principles. Note that, because the stimuli in statistical learning experiments are not generated by rule, they may constitute an example of a sequential learning task, but not a procedural learning task, depending on the definition of the procedural system one adopts.
positive learning outcomes are not consistent with the procedural learning deficit hypothesis, because learning was implicit and stimuli were rule governed. However, it is possible that prosodic patterns are somehow different from patterns in segmental phonology, as well as morphology and syntax. Therefore, it would be helpful to identify another case of phonological pattern learning that does not involve detecting sequential dependencies and on which children or adults with DLD can succeed at the same level as peers with TD.

To that end, the current research employs a segmental sound pattern (one involving consonants and/or vowels) that does not require sequential processing. Because this is a rule-generated, implicitly learned pattern, the procedural learning deficit hypothesis predicts poorer performance from adults with DLD than from adults with TD. In contrast, the sequential pattern learning deficit hypothesis predicts that, in this artificial grammar learning experiment, adults with DLD will show significant learning that does not differ from that seen in adults with TD. In our stimuli, participants listen to a set of familiarization nonwords that are in the form of “Consonant1, Vowel1, Consonant2, Vowel2” (C1V1C2V2) strings. Each nonword must contain at least two out of the following three features: C1 is voiced (C1 voiced), C2 is voiced (C2 voiced), and V1 is front (V1 front). Because not all words must contain the same features (e.g., one word can have C1 voiced and C2 voiced, and another can have C2 voiced and V1 front), this type of pattern—originally drawn from the visual domain—is often referred to as having a Family Resemblance structure (Moreton & Pater, 2012; Shepard et al., 1961).

Family Resemblance patterns have been attested in phonological and morphological systems across human languages (Moreton & Pater, 2012; Moreton et al., 2017). For example, parts of the English irregular past tense system can be described as a Family Resemblance pattern: Irregular verbs that contain /l/ before a velar nasal (e.g., “ring,” “drink,” “swing”) become past tense by undergoing a vowel change (e.g., “rang,” “drank,” “swung”) in the past tense. Verbs that share a subset of these properties (/l/, velar, nasal) also participate in the irregular past tense vowel change (e.g., “swim” to “swam,” “begin” to “began,” “hang” to “hung,” “dig” to “dug”; Bybee & Modér, 1983). Adults have been shown to readily learn Family Resemblance patterns in artificial grammar learning experiments (Gerken et al., 2019; Moreton & Pater, 2012; Moreton et al., 2017).

The Family Resemblance pattern can be described as a set of sequential dependencies. For example, if C1 is voiceless, then C2 is voiced and V1 is front, or if C1 is voiced, then either C2 is voiced or V1 is front, or both. However, this pattern can also be learned without reference to sequential dependencies via phonological feature activation (e.g., Moreton & Pater, 2012). The kind of feature activation that we have in mind can be seen in auditory word-form priming studies in which one word with featural but not full segmental overlap with a target word primes that target (e.g., Goldinger et al., 1989). For example, presenting the auditory word “bull” causes a faster lexical decision for the target word “veer.” Note that “bull” and “veer” have no segments in common, but they do share place of articulation and voicing on C1, contain back vowels, and share manner of articulation and voicing on C2. One way that priming between “bull” and “veer” might occur is that, when a listener hears the word “bull,” the features of that word are activated, which, in turn, activate the same features in other words and thereby activate those words. Thus, when “veer” is heard, it is already weakly activated, which allows a faster response than if the priming word had been featurally unrelated.

The latter example pertains to feature activation during word-form priming. However, what about learning a phonological pattern like the Family Resemblance pattern? Such patterns must be induced over a set of familiarization words followed by a generalization test in which the participant has to determine if a new word fits the pattern. As in the example described above, the familiarization words all share at least two out of the three features: C1 voiced, V1 front, and C2 voiced. As a listener who engages in lexical processing hears the familiarization words, the phonological features in those words are activated. Of course, features in addition to the relevant ones (e.g., place of articulation, vowel height) will be activated as well, but across the set of familiarization words, if these other irrelevant features are randomly distributed, the three relevant features (C1 voiced, V1 front, and C2 voiced) will be the most active. At test, the generalization words can be judged based on how many of the three most active features they contain. Words that contain none or one of the active features are more likely to be rejected as fitting the familiarization pattern, whereas words that contain two or three active features are more likely to be accepted. Thus, the Family Resemblance pattern under consideration here is logically learnable via feature activation.

There is evidence that children and adults with DLD show relatively normal patterns of feature activation. The most direct kind of evidence concerns priming: Children with DLD show phonological priming effects with both phonologically related words (Seiger-Gardner & Schwartz, 2008) and phonologically related nonwords (Brooks et al., 2015). Another type of evidence concerns factors that influence nonword repetition (for a review of many factors, see Szewczyk et al., 2018). As noted above, children and adults with DLD show particular difficulty with nonword repetition compared with their peers with TD; this is not surprising, since nonword repetition is deeply sequential. However, nonwords share characteristics with words, and there is evidence that learners with DLD are influenced by featural overlap between nonwords and known words. For example, children with DLD and children with TD benefit from nonwords that are more similar to words in their language (Archibald & Gathercole, 2006; Graf Estes et al., 2007). When they make errors in nonword repetition, children with DLD and children with TD both generally substitute more frequently occurring phonemes for less frequently occurring phonemes, and their productions tend to be more phonotactically probable than the targets (Burke & Coady, 2015). Children with DLD produce nonwords with high
phonotactic frequency more accurately than those with low phonotactic frequency (Coady et al., 2010; Munson et al., 2005). Munson et al. (2005) showed that children with DLD were actually more influenced by phonotactic frequency than were their age-matched, but not their vocabulary-matched, peers. Finally, one study employing a lexical decision measure also showed that children with DLD were more influenced by the phonotactic probability of nonwords than were their peers with TD (Quémart & Maillart, 2016). We will return to the question of sound-based lexical representations in DLD in the discussion, but for the present purpose, there appears to be sufficient evidence that adults with DLD may well be able to use feature activation to discern featural patterns among auditory nonwords.

Now, consider a phonological pattern that is, on the surface, very similar to the Family Resemblance pattern. In one example of this phonological pattern, all nonwords are C1V1C2V2 strings in which if C1 is voiced, then C2 is voiceless. Now, consider what happens if the verb stem ends in a voiceless segment other than /t/, OR add /d/ if the verb stem ends in /t/ or /d/. However, while adults readily learn Family Resemblance patterns in artificial grammar studies, they fail to learn the OR pattern (Gerken et al., 2019; Moreton & Pater, 2012; Moreton et al., 2017). In contrast, 11-month-olds across several studies show robust learning of the OR pattern (Gerken & Knight, 2015; Gerken & Quam, 2017; Gerken et al., 2019). We will offer an explanation for the developmental differences observed in OR pattern learning in the discussion. For the moment, however, it is important to note that the procedural learning deficit hypothesis predicts that, because both the Family Resemblance and OR patterns are learned implicitly by the procedural system, neither should be learned by adults with DLD. In contrast, the sequential pattern learning deficit hypothesis predicts differential learning of the Family Resemblance and OR patterns by both adults with DLD and adults with TD. The two groups should show equal and robust learning of the Family Resemblance pattern and no learning of the OR pattern.

In summary, children and adults with DLD have difficulty learning some syntactic and phonological components of natural language. They also perform more poorly in most statistical learning and artificial grammar learning tasks. As discussed above, two hypotheses have been proposed to account for this array of observations: the procedural learning deficit hypothesis (Ullman et al., 2020; Ullman & Pierpont, 2005) and the sequential pattern learning deficit hypothesis (Benham et al., 2018; Goffman & Gerken, 2020; Hsu & Bishop, 2014). One way to differentiate these hypotheses is to identify a type of artificial grammar in which the stimuli are generated by a set of rules or principles and that can be learned implicitly but do not involve implicitly tracking sequential dependencies among elements in a string. We contend that phonological Family Resemblance patterns constitute such an artificial grammar.

The procedural learning deficit hypothesis predicts poor performance on implicitly learned, rule-governed patterns, including the Family Resemblance pattern. Therefore, learning this pattern should be more difficult for adults with DLD relative to their peers with TD. In contrast, the sequential pattern learning deficit hypothesis implicates learning only of sequential patterns. Since the Family Resemblance pattern can be learned from activating the set of features in the word and without detecting sequential patterns among the features, this hypothesis predicts that adults with DLD will show significant learning of a phonological Family Resemblance pattern and that their level of learning should not differ from that of adults with TD.

The experiment reported below also examines learning of a phonological OR pattern, which logically requires noting sequential dependencies. The OR pattern was included because the sequential pattern learning deficit hypothesis, but not the procedural learning deficit hypothesis, predicts better performance by adults with DLD on the Family Resemblance pattern than on the OR pattern.

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Method

The main goal of the experiment presented here was to determine if adults with DLD show significant learning of a phonological Family Resemblance pattern and show learning at the same level as adults with TD. Additional goals were to replicate previous findings with adults with TD that demonstrated differential learning of a Family Resemblance pattern and an OR pattern and to determine if, contrary to the procedural learning deficit hypothesis, adults with DLD also show better learning of Family Resemblance than OR patterns.

Participants

Eighty adult college students (21 men, 40 with DLD), ranging in age from 18 to 26 years, participated in the experiment for course credit. All participants indicated that English was their native language. Forty participants were familiarized with a Family Resemblance pattern, and 40 were familiarized with an OR pattern. For each phonological pattern, half of the participants were diagnosed as having DLD, and half were diagnosed as typically developing. Participants were assigned, in alternating order, to each condition until the conditions were filled with 20 subjects each.

The referral pool consisted of individuals from the Department of Psychology undergraduate experiment volunteer pool and from a campus program that provides services to students with language and learning disabilities. Members of the DLD group met the definition of the CATALISE group (Bishop et al., 2016) for DLD as having a language disorder not due to another biological etiology. This is also consistent with how specific language impairment has been operationally defined in the recent literature (see Nitido & Plante, 2020, for a discussion) and with evidence supporting the notion that DLD represents a continuum of behavior rather than a dichotomy based on IQ cut-points (Lancaster & Camarata, 2019). Two individuals with DLD and two in the TD group also reported a diagnosis of attention-deficit/hyperactivity disorder. However, the presence of this condition does not appear to impact the severity of DLD (Redmond et al., 2015). These four individuals had all been assigned to the OR condition. Participants also self-reported an absence of other disorders (i.e., sensory impairment, other neurological disorders).

All participants passed a pure-tone hearing screening and scored above 75 on the Test of Nonverbal Intelligence–Fourth Edition (Brown et al., 2010). This test is scaled so that the normative test mean is 100 and the standard deviation is 15. Language status was determined using the procedures of Fidler et al. (2011), in which performance on a battery of three measures was weighted and the weighted score was compared to a validated cut-point that maximized sensitivity (80%) and specificity (87%) for the classification of individuals as having DLD or typical language skills. The weighting is scaled to a mean of zero, with positive scores corresponding to positive for DLD and negative scores being consistent with typical performance for adults without the disorder. In addition, the Broad Reading subscale of the Woodcock–Johnson Psychoeducational Battery–Third Edition (Woodcock et al., 2001) was given to document reading levels, an additional language skill often impaired in the adult population. In particular, letter–word identification taps decoding, a phonological skill that is often implicated in adults with DLD. Finally, participants completed a nonword repetition task, as this measure specifically taps phonological skills. Nonwords were largely taken from Kamhi and Catts (1986), with five additional four-syllable words added to their original list of 15 nonwords. Table 1 displays the test scores for each set of participants in each condition.

Materials

Materials were C1V1C2V2 nonwords that were created using the schematic shown in Table 2. C1s were “b, g, v, z” (voiced) and “p, k, f, s” (voiceless). C2s were “b, d, z, v” (voiced) and “p, t, f, s” (voiceless). V1s were “r, i” (front) and “o, u” (back). Crossing the eight C1s, eight C2s, and four V1s, with the provision that the same consonant could not occur in C1 and C2, yielded 192 consonant–vowel–consonant–vowel nonwords. The voicing of the first and second consonants and the frontness/backness of the first vowel were manipulated to generate eight different word templates (see Table 2), of which four were consistent with the Family Resemblance pattern and four were consistent with the OR pattern. Note, in Table 2, that the Family Resemblance and OR patterns overlap for words with two voiced consonants. Therefore, two of the word templates were consistent with both patterns (two voiced consonants), two were consistent with only the Family Resemblance pattern (one voiced consonant and a front vowel), two were consistent with only the OR pattern (two voiceless consonants), and two were not compatible with either pattern (one voiced consonant and a back vowel).

Test Words

From the set of 192 words, 32 (four from each of the eight word templates shown in Table 2) were selected for test words. All of the test words had labial consonants (b, p, f, v) in both C1 and C2 positions, because the stimuli were designed to also be used in a production experiment in which lip movements are monitored using articulatory motion capture technology (not reported here). Half (16) of the test words were consistent with the Family Resemblance pattern, and a partially overlapping set of 16 were consistent with the OR pattern (see Table 2). One half of the test words (those with two voiced consonants) were the same for the Family Resemblance and OR conditions. During pilot testing, we eliminated two test words because they were similar or identical to English words, leaving 30 test words in total.

Familiarization Words

Familiarization stimuli were constructed from the pool of consonant–vowel–consonant–vowel nonwords, excluding test words. After elimination of those words that were similar or identical to English or common Spanish words,
two familiarization lists of 76 words each were created, one containing a randomly ordered set of words that are consistent with the Family Resemblance pattern and another containing a randomly ordered set of words that are consistent with the OR pattern. One half of the familiarization words (those with two voiced consonants) were the same for the Family Resemblance and OR conditions. Pauses of 250 ms were placed between words in each list.

**Procedure**

Stimuli were delivered via computer using DirectRT software. Participants were told that they would hear some words and that, later, they would be tested on what they had learned about these types of words just by hearing the words. Critically, they were not given any instruction on what to listen for in the set of words, only that they should listen. After the familiarization phase, participants were asked to respond, via key press, to test words. They were told that the words they had heard conformed to a set of rules and were asked to respond “yes” if the test word also conformed to these rules or “no” if it did not. All test items reflected generalization of the “rules” of the familiarization set to new items. Participants did not receive feedback concerning the accuracy of their responses.

**Results**

A $2 \times 2 \times 2$ Diagnosis (DLD vs. TD) × Pattern Type (Family Resemblance vs. OR) × Consistency (consistent vs. inconsistent with the familiarization pattern) analysis of variance was performed on the mean number of “accept” (“yes” the test item was a member of familiarization words) and “reject” (“no” if it did not) test items. All test items consisted with the familiarization pattern) analysis of variance was performed on the mean number of “accept” (“yes” the test item was a member of familiarization words) and “reject” (“no” if it did not) test items. All test items conformed with the familiarization pattern. One half of the familiarization words (those with two voiced consonants) were the same for the Family Resemblance and OR conditions. Pauses of 250 ms were placed between words in each list.

**Table 1.** Test scores for participant groups in each experimental condition.

<table>
<thead>
<tr>
<th>Test</th>
<th>Family Resemblance condition</th>
<th>OR condition</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>TD</td>
<td>DLD</td>
</tr>
<tr>
<td></td>
<td>M (SD)</td>
<td>Range</td>
</tr>
<tr>
<td>TONI-IV</td>
<td>94.9 (7.8)</td>
<td>84 to 110</td>
</tr>
<tr>
<td>Language Identification Battery</td>
<td>-0.9 (0.7)</td>
<td>-2.08 to -0.02</td>
</tr>
<tr>
<td>W-J Broad Reading</td>
<td>111.2 (12.5)</td>
<td>85 to 113</td>
</tr>
<tr>
<td>W-J Passage Comprehension</td>
<td>104.3 (9.7)</td>
<td>86 to 118</td>
</tr>
<tr>
<td>W-J Letter–Word Identification</td>
<td>102.1 (7.7)</td>
<td>93 to 133</td>
</tr>
<tr>
<td>W-J Reading Fluency</td>
<td>113.4 (13.0)</td>
<td>77 to 140</td>
</tr>
<tr>
<td>Nonword Repetition</td>
<td>15.2 (2.9)</td>
<td>11 to 20</td>
</tr>
</tbody>
</table>

Note. The Test of Nonverbal Intelligence—Fourth Edition (TONI-IV) and the Woodcock–Johnson (W-J) subtests and Broad Reading subscale have a normative mean of 100 and an SD of 15. The Language Identification Battery (Fidler et al., 2011) produces weighted scores, with positive numbers (> 0) indicating developmental language disorder (DLD) status and negative numbers indicating typical language development (TD) status. The Nonword Repetition task scores are number correct out of 20 total.

**Table 2.** Schematic of eight word templates used in the experiment (from Gerken et al., 2019. Copyright 2019 Taylor & Francis. Adapted with permission.).

<table>
<thead>
<tr>
<th>Stimulus description</th>
<th>Short description used in Figures 2 and 3 (C indicates a voiced consonant, and V indicates a front vowel)</th>
<th>Consistent with the Family Resemblance pattern?</th>
<th>Consistent with the OR pattern?</th>
</tr>
</thead>
<tbody>
<tr>
<td>+ front –</td>
<td>CV</td>
<td>yes</td>
<td>no</td>
</tr>
<tr>
<td>+ back +</td>
<td>CC</td>
<td>yes</td>
<td>yes</td>
</tr>
<tr>
<td>– front +</td>
<td>VC</td>
<td>yes</td>
<td>no</td>
</tr>
<tr>
<td>+ front +</td>
<td>CVC</td>
<td>yes</td>
<td>yes</td>
</tr>
<tr>
<td>+ back –</td>
<td>C2</td>
<td>no</td>
<td>no</td>
</tr>
<tr>
<td>– front –</td>
<td>V</td>
<td>no</td>
<td>yes</td>
</tr>
<tr>
<td>– back +</td>
<td>C1</td>
<td>no</td>
<td>no</td>
</tr>
<tr>
<td>– back –</td>
<td>None</td>
<td>no</td>
<td>yes</td>
</tr>
</tbody>
</table>

Note. For C₁ and C₂, “+” indicates voiced, and “−” indicates voiceless. The Family Resemblance pattern requires that at least two of these three features must be present: C₁ voiced, C₂ voiced, and V front (two above). The OR pattern requires that C₁ and C₂ must have the same voicing.
sequential pattern learning deficit hypothesis, there was a significant Consistency × Pattern Type interaction, $F(1, 76) = 22.14, p < .0001, \eta^2_p = .21$. The nature of this interaction is clear in Figure 1, where the “consistent minus inconsistent” acceptance difference is larger for the Family Resemblance group than for the OR group, and this is true for participants with DLD and participants with TD. For the latter group, the difference in acceptance for consistent versus inconsistent test items is near chance (0). None of the other interactions approached significance (Diagnosis × Consistency, $F < 1$; Diagnosis × Pattern Type, $F < 1$; three-way interaction, $F < 1$).

The sequential pattern learning deficit hypothesis predicted that adults with DLD would learn the Family Resemblance pattern and that they would do so at the same level as adults with TD. Therefore, although the analysis of variance showed no interactions with diagnosis, separate $t$ tests were performed on the four groups (DLD Family Resemblance, TD Family Resemblance, DLD OR, and TD OR) independently. Adults in the DLD Family Resemblance group significantly differentiated consistent from inconsistent test items, $M (SD)$: consistent = 11.45 (3.97), inconsistent = 7.72 (4.36); $t(19) = 5.12, p < .0001, \text{Cohen's } d = 1.15$. Replicating previous findings (e.g., Gerken et al., 2019), adults in the TD Family Resemblance group significantly differentiated consistent from inconsistent test items, $M (SD)$: consistent = 11.28 (2.95), inconsistent = 7.57 (2.81); $t(19) = 4.75, p < .0002, \text{Cohen's } d = 1.06$. Both the DLD and TD groups who were familiarized with the Family Resemblance pattern showed large to very large effect sizes for discriminating consistent from inconsistent test items.

Unlike the robust discrimination of consistent versus inconsistent test items seen for the Family Resemblance pattern, the DLD group accepted only one (CC; see Figure 3) of the four consistent test item types at a higher rate than for all four inconsistent test item types. The TD group’s most accepted test item type (CV; see Figure 3) was actually inconsistent with the familiarization words. As noted above, both groups showed a near-zero difference between consistent and inconsistent test items (see Figure 1).

Discussion

The results from the experiment were remarkably clean. Briefly summarizing, adults with DLD showed significant learning of the Family Resemblance pattern, and they did
Figure 2. Mean acceptance rates (SE) for test items that were consistent (black) versus inconsistent (gray) with Family Resemblance pattern familiarization items. For more information about the test items, see Table 2. C = voiced consonant; DLD = developmental language disorder; TD = typical language development; V = front vowel.

Figure 3. Mean acceptance rates (SE) for test items that were consistent (black) versus inconsistent (gray) with OR pattern familiarization items. For more information about the test items, see Table 2. C = voiced consonant; DLD = developmental language disorder; TD = typical language development; V = front vowel.
so at a comparable level as adults with TD, as evidenced by similar means for consistent versus inconsistent test items across the two diagnosis groups, by similar large to very large effect sizes for t tests comparing consistent and inconsistent test items, and by a higher acceptance rate for all consistent test items than for all four inconsistent test items. The strong performance of adults with DLD on the Family Resemblance pattern is consistent with their strong performance in an earlier study of stress pattern learning (Bahl et al., 2009). Thus, the current study, coupled with the earlier one on stress pattern learning, suggests that adults with DLD show a varied topography of strengths and weaknesses in artificial grammar learning studies—one that does not cleave neatly to the procedural–declarative dichotomy. We will return to the causes of these strengths and weaknesses below.

In contrast to strong performance on learning the Family Resemblance pattern, adults with DLD and adults with TD who were familiarized with the OR pattern failed to show learning and showed significantly poorer performance than adults who were familiarized with the Family Resemblance pattern. The failure of adults with TD to learn the OR pattern replicates previous studies (Gerken et al., 2019; Moreton & Pater, 2012; Moreton et al., 2017). The differential performance of adults with DLD on the Family Resemblance versus OR patterns conceptually replicates other research on artificial grammar learning by these adults (e.g., stress pattern learning vs. sequential dependency learning). Below, we discuss why adults with TD and adults with DLD might have failed to learn the OR pattern and if they failed for the same reason.

**Implications for Accounts of DLD**

Although there are a number of accounts of DLD, we focused here on two: the procedural learning deficit hypothesis (e.g., Ullman et al., 2020) and the sequential pattern learning deficit hypothesis (Benham et al., 2018; Goffman & Gerken, 2020; Hsu & Bishop, 2014). These two accounts both treat the morphosyntactic and phonological weaknesses observed in DLD as arising from a single underlying mechanism, and both propose that this underlying mechanism is not specific to language. Given the similarities between these two accounts, the main goal of the research presented here was to compare the performance of adults with DLD and adults with TD on an artificial grammar pattern that can be learned without requiring the detection of contingency relations in a sequence; this pattern should be learnable on the sequential pattern learning deficit hypothesis, but not on the procedural learning deficit hypothesis. In the introduction, we described one type of previous artificial grammar learning study involving stress assignment principles that also does not require detecting a contingency in a sequential pattern (Bahl et al., 2009; Plante et al., 2010). In two studies using this grammar, adults and children with DLD performed on par with their peers with TD. The comparable behavior of participants with DLD and participants with TD in these studies stands in contrast to the more usual finding in statistical learning and artificial grammar learning experiments, in which participants with DLD significantly underperform compared to their peers with TD. However, as we noted in the introduction, stress assignment may be unique in some way, raising the need for a more typical artificial grammar learning study that might differentiate the two hypotheses under consideration.

The results from participants in the Family Resemblance group supported the predictions of the sequential pattern learning deficit hypothesis, but not the procedural learning deficit hypothesis. We suggest that the Family Resemblance pattern can be learned via feature activation and, therefore, does not require detecting contingencies between elements in a sequence. On the account proposed here, participants with DLD and participants with TD both learned the Family Resemblance pattern via feature activation. However, as discussed in the introduction, the OR pattern cannot be learned via feature activation. Rather, it requires learners to detect sequential dependencies between C₁ and C₃. In contrast with the sequential pattern learning deficit hypothesis, the procedural learning deficit hypothesis predicted poorer performance by our participants with DLD than by those with TD on both the Family Resemblance and OR patterns, because both patterns are rule generated and both are learned implicitly.

We contend here that the Family Resemblance pattern falls squarely in the domain of procedural learning as that construct has been employed to explain DLD (e.g., Ullman et al., 2020), as well as how it has been used in related literature. Indeed, on a somewhat different view of the procedural learning system than the one espoused by Ullman et al. (2020), only the Family Resemblance pattern and not the OR pattern is learned via the procedural system, because the former involves integrating over stimulus dimensions (e.g., Ashby & Maddox, 2005; Smith et al., 2012). The Family Resemblance pattern is generated by rule and is learned implicitly and, therefore, fits the definition of a procedural task given by Ullman and Pierpont (2005). One possible objection to our contention that the Family Resemblance pattern should be learned by the procedural system as it is defined under the procedural learning deficit hypothesis concerns our suggestion that the Family Resemblance pattern is learned via feature activation within the form-based lexicon. Thus, it might be possible for the proponents of the procedural learning deficit hypothesis to claim that, because feature activation occurs in the lexicon, it is outside the domain of procedural learning.² We offer three arguments against this claim. First, our interpretation concerns the word-form lexicon, not the semantic lexicon or arbitrary associations between forms and referents. Only the latter two are thought to implicate declarative memory (Ullman & Pierpont, 2005). Second, adults appear to have implicit (and not explicit) access to various grammar-governed regularities of the word-form lexicon that seem to reflect just

²We are grateful to an anonymous reviewer for pointing out this issue.
the sort of rules that the procedural learning account of DLD was created to address. For example, neither /bw/ nor /dl/ occurs at the beginning of English words, yet /bw/ is an accidental gap, and /dl/ is a grammatically driven gap, according to the Obligatory Contour Principle (e.g., Frisch, 2004). English-speaking adults distinguish these forms, treating the accidental gap as more acceptable than the grammatical gap (e.g., Moreton, 2002). Such results suggest that implicit, abstract grammatical principles of the sort that are the focus of the procedural deficit account can arise from word forms in the lexicon. Finally, the learning of stress assignment principles described in the introduction also appears to fit the definition of procedural learning. They are rule governed, implicit, and do not involve memorized, idiosyncratic mappings (Ullman & Pierpont, 2005). Yet, they are learned by both children and adults with DLD at the same level as their peers with TD, and they do not appear to be learned via feature activation. Thus, the emerging range of observations about when children and adults with DLD succeed versus fail to learn various linguistic patterns appears to be more consistent with the sequential pattern deficit hypothesis than with the procedural deficit hypothesis. It is in the sequential dependencies that the learner breaks down.

**Implications for Typical and Atypical Language Development**

The current study replicates previous work in which adults are not able to learn an OR pattern in the lab (Gerken et al., 2019; Moreton & Pater, 2012; Moreton et al., 2017). In contrast, as noted in the introduction, 11-month-olds are readily able to learn the OR pattern (Gerken & Knight, 2015; Gerken & Quam, 2017; Gerken et al., 2019), even from just four familiarization words (Gerken & Knight, 2015). One study directly compared adult and infant learning of the same OR pattern (the OR pattern employed here); 11-month-olds learned the pattern, but adults did not (Gerken et al., 2019). Elsewhere, we have already noted that children and adults with DLD show phonological priming effects (Brooks et al., 2004; Archibald et al., 2013; Benham et al., 2018). Yet, the sequential pattern learning deficit hypothesis, which was supported by the experiment presented here, depends on adults with DLD having a relatively normally organized form-based lexicon. It is in this lexicon that features of incoming nonwords are activated, allowing nonsequential, feature-based patterns, such as the Family Resemblance pattern, to be detected. How can we reconcile apparent phonological deficits with relatively normally organized lexicon? We have already noted that children and adults with DLD show phonological priming effects (Brooks et al., 2004; Seiger-Gardner & Schwartz, 2008). They are also influenced by many of the same factors as their peers with TD in nonword repetition, new word learning, and lexical decision. These factors include word-likeliness (Archibald & Gathercole, 2006; Graf Estes et al., 2007), phoneme frequency, and phonotactic probability (Burke & Coady, 2015; Coady et al., 2010; Munson et al., 2005; Quémart & Maillart, 2016). Nevertheless, there is ample evidence that the phonological representations of children and adults with DLD are somehow weaker or less well specified than the representations of their peers with TD (Alt & Plante, 2006; Archibald et al., 2013; Coady & Evans, 2008; Dollaghan, 1998; Edwards & Lahay, 1998). These less robust representations might be sufficient for feature activation of the type implicated by the sequential pattern learning deficit hypothesis and the experiment presented here. However, these representations might not be sufficient for tasks such as word learning or lexical decision, in which an actual item in the form-based lexicon must be uniquely accessed and integrated with its counterpart in the semantic lexicon (e.g.,

The developmental account offered here raises an important question about whether young children (or infants) with DLD could learn the OR pattern. There are two possibilities. On the sequential pattern learning deficit hypothesis, the OR pattern reflects exactly the type of contingent sequence learning that is weak in DLD. Therefore, we expect that young children with DLD would not be able to learn this pattern as well as their peers with TD. If DLD is characterized by a specific difficulty with sequential patterns, the adults with DLD in our study would have never been able to learn the OR pattern because of their weakness in sequential processing. In contrast, adults with TD were able to learn OR patterns as young children until their lexical processing bias became too strong. The second possibility is that adults with DLD were, at some point in their infancy or childhood, able to learn the OR pattern and have lost that ability, due to a developmentally increasing bias, to engage in lexical processing of speech that is comparable to the time course over which this happens with typically developing children. If this is the case, young children with DLD may be able to learn the OR pattern early in development. We are testing this possibility in our labs.

**What is the Nature of the Lexicon in DLD?**

As discussed in the introduction, phonological deficits are an increasingly acknowledged component of DLD (e.g., Alt et al., 2004; Archibald et al., 2013; Benham et al., 2018). Yet, the sequential pattern learning deficit hypothesis, which was supported by the experiment presented here, depends on adults with DLD having a relatively normally organized form-based lexicon. It is in this lexicon that features of incoming nonwords are activated, allowing nonsequential, feature-based patterns, such as the Family Resemblance pattern, to be detected. How can we reconcile apparent phonological deficits with a relatively normally organized lexicon? We have already noted that children and adults with DLD show phonological priming effects (Brooks et al., 2004; Seiger-Gardner & Schwartz, 2008). They are also influenced by many of the same factors as their peers with TD in nonword repetition, new word learning, and lexical decision. These factors include word-likeliness (Archibald & Gathercole, 2006; Graf Estes et al., 2007), phoneme frequency, and phonotactic probability (Burke & Coady, 2015; Coady et al., 2010; Munson et al., 2005; Quémart & Maillart, 2016). Nevertheless, there is ample evidence that the phonological representations of children and adults with DLD are somehow weaker or less well specified than the representations of their peers with TD (Alt & Plante, 2006; Archibald et al., 2013; Coady & Evans, 2008; Dollaghan, 1998; Edwards & Lahay, 1998). These less robust representations might be sufficient for feature activation of the type implicated by the sequential pattern learning deficit hypothesis and the experiment presented here. However, these representations might not be sufficient for tasks such as word learning or lexical decision, in which an actual item in the form-based lexicon must be uniquely accessed and integrated with its counterpart in the semantic lexicon (e.g.,
Jones & Brandt, 2018; Storkel et al., 2010). In short, despite having weaker phonological representations, children and adults with DLD may be able to use their form-based lexicons to identify some types of phonological patterns, some of which have parallels in morphosyntax, and thereby compensate for their difficulties with sequential pattern learning of the type used here. Clearly, more research on the specific nature of word-form representations is needed before we can predict how much compensation can be attained.

Conclusions

The fact that adults with DLD showed very robust learning of an artificial grammar involving a Family Resemblance phonological pattern supported the sequential pattern learning deficit hypothesis over the procedural deficit hypothesis. This finding suggests that any adequate account of DLD needs to view artificial grammar learning as a multifactorial problem that can showcase both strengths and weaknesses. Hopefully, this observation will lead to new, more nuanced approaches to artificial grammar and statistical learning by children and adults with DLD.

Author Contributions

LouAnn Gerken: Conceptualization (Equal), Formal analysis (Lead), Funding acquisition (Supporting), Writing - original draft (Lead). Elena Plante: Investigation (Equal), Methodology (Equal), Project administration (Lead), Writing - review & editing (Equal). Lisa Goffman: Conceptualization (Equal), Funding acquisition (Supporting), Methodology (Equal), Writing - review & editing (Equal).

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