Morphological cues vs. number of nominals in learning verb types in Turkish: The syntactic bootstrapping mechanism revisited

A. Engin Ural and Deniz Yuret
Koç University, Istanbul, Turkey

F. Nihan Ketrez
Yale University, New Haven, CT, USA

Dilara Koçbaş and Aylin C. Küntay
Koç University, Istanbul, Turkey

The syntactic bootstrapping mechanism of verb learning was evaluated against child-directed speech in Turkish, a language with rich morphology, nominal ellipsis and free word order. Machine-learning algorithms were run on transcribed caregiver speech directed to two Turkish learners (one hour every two weeks between 0:9 to 1:10) of different socioeconomic backgrounds. We found that the number of nominals in child-directed utterances plays a small, but significant, role in classifying transitive and intransitive verbs. Further, we found that accusative morphology on the noun is a strong cue in clustering verb types. We also found that verbal morphology (past tense and bareness of verbs) is useful in distinguishing between different subtypes of intransitive verbs. These results suggest that syntactic bootstrapping mechanisms should be extended to include morphological cues to verb learning in morphologically rich languages.

Correspondence should be addressed to Aylin C. Küntay, Department of Psychology, Koç University, Rumeli Feneri Yolu, Sarıyer, Istanbul 34450, Turkey. E-mail: akuntay@ku.edu.tr

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INTRODUCTION

Syntactic bootstrapping is one of the learning mechanisms proposed to explain how children determine verb meanings from input. According to this account, children keep track of the number of arguments across utterances to derive general meanings such as transitive or intransitive verb types (Fisher, Hall, Rakowitz, & Gleitman, 1994; Gleitman, 1990; Landau & Gleitman, 1985; Naigles, Gleitman, & Gleitman, 1993; Naigles & Swensen, 2006). For example, a two-argument frame such as ‘the dog is verb-ing the cat’ implies a transitive act, while a one-argument utterance such as ‘the dog is verb-ing’ is likely to be construed as an intransitive act. When language learners encounter a verb mostly in one-argument frames, they would conclude that this verb is an intransitive verb.

There is substantial evidence that 2-year-old children make use of argument number and argument placement information in syntactic frames to derive verb meanings (e.g., Hirsh-Pasek, Golinkoff, & Naigles, 1996; Naigles, 1998; see also Lidz, Gleitman, & Gleitman, 2003 for Kannada-speaking children). Gertner, Fisher, and Eisengart (2006) found that children as young as 21 months old use argument order information appropriately to interpret who is doing what to whom in transitive constructions containing nonsense verbs. However, most research on syntactic bootstrapping has involved English learners, and focused on the argument composition of the construction rather than nominal or verbal morphology (Naigles & Swensen, 2006).

Whether syntactic bootstrapping is a plausible mechanism of language learning cross-linguistically is a topic of current research interest. Given the variation in grammatical devices of the world’s languages, the English language, featuring a strict word order and no nominal ellipsis, reflects an ideal case of correspondences between argument number and verb types. Several criticisms of the syntactic bootstrapping approach have questioned the generalisability of the syntactic bootstrapping mechanism to languages with non-English-like grammatical properties (e.g., Allen, 2007; Bowerman & Brown, 2007; Goldberg, 2006; Rispoli, 1995; Wilkins, 2007).

In testing the cross-linguistic viability of the syntactic bootstrapping approach with Mandarin learners, Lee and Naigles (2005) examined a corpus of 6,088 utterances directed to children acquiring Mandarin, which allows argument ellipsis and flexible word order, similarly to Turkish. A total of 7,884 tokens of 60 most frequent verbs were manually tagged to reveal that learners hear post-verbal noun phrases more frequently with transitive
verbs than intransitive ones, suggesting that the number of arguments surrounding a verb can provide useful information for Mandarin children to classify verb types. Göksun, Küntay, and Naigles (2008) carried out an experimental sentence act-out study with child and adult speakers of Turkish using two-argument and one-argument constructions and toy animal props. They investigated whether Turkish speakers construe constructions more causatively depending on the number of explicit arguments in the sentence, the presence of accusative case marking, and the verbal causative morpheme. Causative construals were act-outs in which participants made one animal act upon another animal by changing its state or position. The results revealed that the number of arguments in addition to case marking morphology in the construction affected the likelihood of ‘transitive’ causative enactments. These findings render the syntactic bootstrapping proposal plausible for languages with ellipsis and flexible word order.

In this study, we administer machine-learning experiments on child-directed Turkish recorded during family interactions to determine further the cross-linguistic feasibility of the syntactic bootstrapping approaches to early verb learning in Turkish child-directed speech and to explore the value of various morphosyntactic cues to distinguish different subclasses of verbs. Although there were previous automated natural-language processing techniques used to categorise verbs into semantic classes (e.g., Schulte im Walde, 2006; Merlo & Stevenson, 2001), none of these previous attempts tested theories of language development.

Turkish, unlike English and Mandarin, is a language that (a) relies on morphology, not so much on argument ordering, to assign grammatical relations in a clause, and similarly to Mandarin (b) allows extensive argument ellipsis and word order alternations. As Naigles and Swensen (2006) state, experimentation within the syntactic bootstrapping model has mostly been limited to English, and not yet adequately applied to languages with complex morphology. Therefore, Turkish language provides a good testing ground for the syntactic bootstrapping mechanism in a language that mainly employs morphological means for indicating ‘who-does-what-to-whom’.

Turkish is an SOV language, where verbs typically appear at the end of utterances. However, Turkish allows omission of both the subject and the object, and has flexible word order. Being an agglutinating language, Turkish marks case (accusative, dative, locative, ablative, instrumental, genitive), possessive and plurality on nouns, which include pronouns. Verb morphology includes tense/aspect/modality markers, negation, subject agreement and voice morphemes (passive, causative, reflexive, and reciprocal). Verbs can appear in the bare form in the imperative mood.

The current study addresses three questions using machine-learning algorithms on child-directed speech: (1) Is the number of nominals in a
sentence informative for classifying transitive and intransitive verb types in Turkish child-directed speech? (2) Does nominal morphology facilitate distinguishing transitive versus intransitive verbs? (3) Does verbal morphology play any role in distinguishing between different subtypes of transitive and intransitive verbs?

**METHOD**

**Dataset preparation**

Child–caregiver interaction was video-recorded at the homes of two female Turkish learners (Irmak and Elif), for one hour every two weeks. Irmak’s parents both had 8 years of education; Elif’s parents both had doctoral degrees. Table 1 provides the basic characteristics of the datasets. The Irmak corpus and the Elif corpus respectively contained 12,276 and 20,687 morphologically coded child-directed utterances, which were the data used by our machine learning programs.

Trained native speakers transcribed and morphologically coded the spoken language from the videotapes using the CHAT transcription format provided by the CHILDES project (MacWhinney, 2000). Below is a sample utterance from the child-directed data and its morphological coding tier.

(1) NAN-CHI: kuş-lar-i-m arı-yo(r)-sun?
%mor: N|kus-PL-ACC-QUE V|ara-PROG-2S
‘Are you looking for the birds?’

The datasets for the two children exhibited remarkable differences as well as similarities. Table 2 shows the most frequent 10 verbs in each child’s data set, 8 of which are common. Even though the total number of utterances was

<table>
<thead>
<tr>
<th>TABLE 1</th>
<th>Characteristics of the databases</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Irmak</td>
</tr>
<tr>
<td>Number of sessions</td>
<td>30</td>
</tr>
<tr>
<td>Start age</td>
<td>0;9,0</td>
</tr>
<tr>
<td>End age</td>
<td>2;0,16</td>
</tr>
<tr>
<td>Total number of utterances</td>
<td>32362</td>
</tr>
<tr>
<td>Child-directed utterances</td>
<td>15781</td>
</tr>
<tr>
<td>Child-directed with morphological analysis</td>
<td>12276</td>
</tr>
<tr>
<td>Number of verb lemma tokens</td>
<td>18162</td>
</tr>
<tr>
<td>Number of unique verb lemmas</td>
<td>601</td>
</tr>
<tr>
<td>Verb lemmas with count &gt;10</td>
<td>124</td>
</tr>
</tbody>
</table>
approximately the same, the number of child-directed utterances is higher in the Elif dataset. The frequency distribution of verbs in the Elif dataset is flatter: The top 10 verbs in the Elif dataset account for 45% of all verb occurrences, whereas in the Irmak data this ratio is 55%. These factors result in more verb types in the Elif dataset above a frequency threshold (170 verbs appearing ≥ 10 times) compared to the Irmak data set (124 verbs appearing ≥ 10 times). These differences can be due to somewhat different language environments created by families of different socio-educational backgrounds. However, 105 verbs were common across the two corpora. From a verb learning perspective, our study suggests that the algorithms work well on both datasets.

**Verb categorisation**

We manually tagged each verb that occurred 10 times or more in our corpora according to their argument structure as unaccusative, unergative, and transitive. Our argument structure categorisation of Turkish verbs was based on a modified version of Ketrez (1999), which adopted an analysis based on Grimshaw’s (1992) Prominence Theory. We used the transitive-intransitive classification as our primary learning target. The transitive category included Grimshaw’s (1992) transitive agentive verbs (xagent (ytheme)) such as aç-‘open’, kır-‘break’, ye-‘eat’, which involve an agent and a theme that was acted upon by the agent, and ditransitive verbs (xagent (ygoal (ztheme))) that include koy-‘put’, ver-‘give’, where a third argument in the form of a goal is included in the event. We also experimented with splitting the intransitive verbs into unaccusatives ((xtheme)) that have only a ‘theme’ argument such as diş-‘fall’ and unergative (xagent) verbs that has an agent as the sole argument, such as koş-‘run’.

<table>
<thead>
<tr>
<th>Root</th>
<th>Gloss</th>
<th>Freq</th>
<th>Root</th>
<th>Gloss</th>
<th>Freq</th>
</tr>
</thead>
<tbody>
<tr>
<td>gel</td>
<td>come</td>
<td>2809</td>
<td>gel</td>
<td>come</td>
<td>1900</td>
</tr>
<tr>
<td>bak</td>
<td>look</td>
<td>2302</td>
<td>yap</td>
<td>do</td>
<td>1848</td>
</tr>
<tr>
<td>al</td>
<td>take</td>
<td>1016</td>
<td>bak</td>
<td>look</td>
<td>1835</td>
</tr>
<tr>
<td>git</td>
<td>go</td>
<td>865</td>
<td>ol</td>
<td>be</td>
<td>1196</td>
</tr>
<tr>
<td>yap</td>
<td>do</td>
<td>857</td>
<td>git</td>
<td>go</td>
<td>742</td>
</tr>
<tr>
<td>de</td>
<td>say</td>
<td>576</td>
<td>iste</td>
<td>want</td>
<td>625</td>
</tr>
<tr>
<td>ol</td>
<td>be</td>
<td>459</td>
<td>al</td>
<td>take</td>
<td>604</td>
</tr>
<tr>
<td>ye</td>
<td>eat</td>
<td>398</td>
<td>koy</td>
<td>put</td>
<td>490</td>
</tr>
<tr>
<td>otur</td>
<td>sit</td>
<td>362</td>
<td>ver</td>
<td>give</td>
<td>444</td>
</tr>
<tr>
<td>ver</td>
<td>give</td>
<td>352</td>
<td>ye</td>
<td>eat</td>
<td>433</td>
</tr>
<tr>
<td>Total</td>
<td></td>
<td>9996</td>
<td></td>
<td></td>
<td>10117</td>
</tr>
</tbody>
</table>
Feature evaluation

To determine if the utterance context of a verb carries sufficient information to correctly classify its argument structure, we used machine learning methods to evaluate different features of the verb context. We considered several different types of noun and verb features that may facilitate identification of the verb argument structures. The choice of this comprehensive set of features was based on features considered important in previous work done on Turkish child language and child-directed speech (e.g., Aksu-Koç & Ketrez, 2003; Aksu-Koç & Slobin, 1985; Ketrez, 1999; Küntay & Slobin, 1996). Table 3 presents the features evaluated under three different subheadings:

- The average number of nominals that occur with the verb in a given utterance.
- The frequency with which the verb is observed with arguments with different types of case markers.
- The frequency with which the verb occurs with different types of inflectional and derivational morphemes.

The WEKA machine learning toolkit (Witten & Frank, 2005) was used to systematically evaluate different subsets of these features and rank them according to their performance in discriminating transitive from intransitive verbs, and within intransitives discriminating unaccusatives from unergatives. We took the distance between two verbs to be the Euclidean distance between their feature vectors. We used the k-means clustering algorithm (Bishop, 2006) to measure the performance of each feature subset considered. The k-means algorithm partitioned the given verbs into k clusters (k = 2 in our case) minimising the distance of each verb to the centroid, or prototype of its cluster. We then compared the resulting cluster assignments with the manually tagged actual categories of the verbs to see how well a particular feature set performs.

The following results focus on the features which proved most useful in providing verb clusters, among those considered in this study and listed in Table 3.

RESULTS AND DISCUSSION

Number of nominals is moderately effective in verb categorisation

We assessed whether the number of nominals that surround a verb in an utterance was a useful cue in distinguishing between transitive and
intransitive verbs. Among nominals we included words that might act as verb arguments such as nouns, pronouns, and some wh-words that had pronominal properties (i.e., ‘who’ and ‘what’). If an utterance contained a single verb, we assigned all the nouns in that utterance to that verb. If an utterance included multiple verbs, we assigned each noun to the nearest verb on its right as Turkish is right-headed. In those cases where there was no verb on the right, we used the nearest verb on the left.
Table 4 presents the mean number of nominals surrounding transitive and intransitive verbs in both datasets. The number of nominals around the transitive verbs is significantly higher than around the intransitive verbs, with small-to-moderate effect sizes for both datasets: Elif, $t(168) = 4.30$, $p < .0001$, Cohen’s $d = 0.66$; Irmak, $t(122) = 4.16$, $p < .0001$, $d = 0.75$.

The number of nominals was used as input to the k-means clustering algorithm to determine the extent to which this feature facilitates the prediction of the transitive-intransitive distinction in child-directed speech. On the basis of a majority class baseline measure, which always predicts the most frequent category (i.e., the transitive), the accuracies for detecting the right verb category would be 57.1% and 61.3% for Elif and Irmak, respectively. The usage of the feature of the number of nominals around the verb would increase these predictions up to 68.2% and 71.8%, only about 10% better than the baseline.

**Nominal morphology strongly signals transitivity**

Next we assessed whether accusative marking was a stronger indicator of the transitive-intransitive distinction. Table 5 presents the proportion of accusative-marked nominals that accompany transitive and intransitive verbs for the Elif and Irmak datasets. The frequency of accusative-marked nominals is significantly higher for transitive verbs compared to intransitive

<table>
<thead>
<tr>
<th></th>
<th>Elif</th>
<th>Irmak</th>
</tr>
</thead>
<tbody>
<tr>
<td>Transitive</td>
<td>.34 (.04)</td>
<td>.27 (.04)</td>
</tr>
<tr>
<td>Intransitive</td>
<td>.02 (.007)</td>
<td>.02 (.008)</td>
</tr>
</tbody>
</table>

The 95% confidence interval of the mean proportions is shown in parentheses.
verbs, with large effect sizes for both datasets: Elif, \( t(168) = 14.02, p < .0001, d = 2.32; \) Irmak, \( t(122) = 9.88, p < .0001, d = 2.01. \)

Using the frequency of accusatives as input to the k-means clustering algorithm, we can distinguish transitives from intransitives correctly at 92% of the time in the Elif dataset and 90% of the time in the Irmak dataset. The accusative morphology on the noun increases the accuracies for detecting the verb category by about 30% compared to majority class baseline predictions (i.e., 57.1% for Elif and 61.3% for Irmak). The results so far cumulatively demonstrate that although number of nominals plays some role in distinguishing between verb classes, the presence or absence of accusative morphology on the noun constitutes a stronger cue.

**Further subcategorisation with verb morphology**

We explored whether verbal morphology would be useful in distinguishing between different classes of verbs. Among the features considered (see Table 3), the past morpheme and lack of verbal morphology on the verb, which is used for imperatives in Turkish, contributed criterial value in subcategorising intransitive verbs into unaccusatives and unergatives.

Unaccusative verbs, such as \( \text{duş-} \)‘fall’, that usually denote states and whose subjects have the ‘theme’ role are practically never used in the bare imperative form, so the frequency of the use of the bare form of the verb distinguishes unaccusatives from unergatives well. Table 6 provides the proportion of unaccusative and unergative verbs that appear with no inflectional morphology in the Elif and Irmak datasets. The frequency of having no morphology on the verb is significantly higher for unergative verbs compared with unaccusative verbs, with large effect sizes for both datasets: Elif, \( t(71) = 4.08, p < .0001, d = 1.03; \) Irmak, \( t(46) = 6.39, p < .0001, d = 2.06. \)

Using the frequency with which verbs are used in their bare form as input to the k-means algorithm, we were able to predict the unaccusative-unergative distinction with 74% accuracy in the Elif dataset, and 83% accuracy in the Irmak dataset. Both numbers are above the baseline of

<table>
<thead>
<tr>
<th>TABLE 6</th>
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<tbody>
<tr>
<td>The proportions of unaccusative and unergative verbs appearing with no verbal morphology in the Elif and Irmak datasets</td>
</tr>
<tr>
<td></td>
</tr>
<tr>
<td>Unaccusative</td>
</tr>
<tr>
<td>Unergative</td>
</tr>
</tbody>
</table>

The 95% confidence interval of the mean proportions is shown in parentheses.
always predicting the most frequent category (58% for Elif and 60% for Irmak).

Table 7 provides the proportion of unaccusative and unergative verbs that appear with past morphology in the Elif and Irmak datasets. The frequency of having past morphology on the verb, which marks the perfective aspect as well as the past tense, is significantly higher for unaccusative verbs compared with unergative verbs, with large effect sizes for both datasets: Elif, \( t(71) = 4.30, \ p < .0001, \ d = 0.99 \); Irmak, \( t(46) = 5.37, \ p < .0001, \ d = 1.48 \). The unaccusative-unergative distinction was predicted with 73% and 64% accuracy for Elif and Irmak datasets respectively and these numbers are again above the baseline of always predicting the most frequent category (58% for Elif and 60% for Irmak).

Unaccusative verbs are mostly used in contexts that describe the final state of the objects (e.g., the toy fell/broke). In this sense, the present finding is not surprising and it is in line with the observation in Aksu-Koç (1988, 1998), who reports that in both child and child-directed speech, achievement verbs that overlap with the unaccusatives in the present study mostly appear with past morphology –DI, denoting completed action and final state of the objects. This correlation between verbal morphology and aspectual properties of verbs in child language has been observed in other languages as well, from as early as the 1970s (Antunicci & Miller, 1976, Bloom, Lifter, & Hafitz, 1980; Clark, 1996).

**CONCLUSION**

Recent work with non-English languages (e.g., Göksun et al., 2008; Lee & Naigles, 2005; Lidz, Gleitman, & Gleitman, 2003) has shown that the number of arguments in a sentence plays a role in distinguishing transitive verbs from intransitive verbs in languages that allow nominal ellipsis. What remains unclear is whether morphological features constitute a stronger cue in morphologically complex languages.

<table>
<thead>
<tr>
<th></th>
<th>Elif</th>
<th>Irmak</th>
</tr>
</thead>
<tbody>
<tr>
<td>Unaccusative</td>
<td>.39 (.08)</td>
<td>.41 (.12)</td>
</tr>
<tr>
<td>Unergative</td>
<td>.19 (.05)</td>
<td>.09 (.05)</td>
</tr>
</tbody>
</table>

The 95% confidence interval of the mean proportions is shown in parentheses.
We have shown that in a language with agglutinative morphology such as Turkish, number of nominals surrounding a verb is not the only cue for determining verb categories. In classifying verb types, morphological cues provide more reliable cues than the number of nominals in speech to two female children of different socioeconomic backgrounds. The presence of accusative case marking in child-directed utterances leads to better clustering of transitive and intransitive verbs compared with the number of nouns. In addition, absence of verbal morphology found in imperatives and past tense morphology appear criterial in partitioning between different types of intransitive verbs. The evidence we have so far from early language production of Turkish children makes these findings not very surprising. The accusative is the first nominal inflection and the past tense is the first verbal inflection to appear in early child language (Aksu-Koç & Ketrez, 2003; Aksu-Koç & Slobin, 1985; Ketrez, 1999, among others). The frequency and the regularity of these morphological cues probably render them highly learnable morphosyntactic devices.

There is nothing in the syntactic bootstrapping account that precludes morphological cues from being considered as cues to verb argument structure, however, the existing literature mostly focuses on the number and semantic types of arguments without seriously considering morphology. The syntactic bootstrapping mechanism would be more viable if the role of morphological features such as case and tense are taken into consideration in the acquisition of verb meanings, at least for morphologically rich languages.

As an example, the Göksun et al. (2008) study with Turkish learners in four groups of preschool ages and adults evaluates both the presence of inflectional morphology and the number of nouns as potential cues to whether verbs are acted out as causative or non-causative events. Causative enactments increased in two-noun constructions and decreased in single-noun constructions. However, the effect of the number of explicit arguments was found to be stronger in children than in adults and less robust than that in English (Naigles, Kuntay, Göksun, & Lee, 2006). Another finding was that the role of accusative morphology in leading to causative act-outs was apparent from the earliest age studied (i.e., 2-year-olds) (see also Slobin & Bever, 1982). In sum, the Göksun et al. (2008) study showed that Turkish children and adults are able to use nominal morphological information, as well as number of arguments, when making conjectures about construction meaning. In this current work, the most potent device we find in child-directed speech to tell apart transitive verbs from intransitive verbs is the accusative casemarking. However, the number of arguments permits classification of verbs by transitivity beyond a baseline as well.

In brief, the two kinds of data in Turkish – experimental act-out studies of sentence comprehension and a corpus analysis of child-directed speech – converge to show that the syntactic bootstrapping mechanism should be
extended as ‘morphosyntactic bootstrapping’ (Göksun et al., 2008) to include morphological cues. More research is needed to determine how different linguistic devices interact in different ways across different languages and at different developmental points of language development (Dittmar, Abbot-Smith, Lieven, & Tomasello, 2008).

REFERENCES


