

What Do Hindi Native Speakers Predict?

INVESTIGATING VERB CLASS, VERB MORPHOLOGY AND WORD ORDER

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Introduction

■ *abhay ne roti*

Abhay ERG bread

- It has been established that human sentence processing involves prediction, i.e. we do not wait for the entire utterance to complete before making sense of the utterance
- Prediction can be thought as some kind of top-down process wherein the upcoming linguistic material becomes activated before it is actually encountered

Introduction

- While it is undisputed that we predict, the exact mechanism of the prediction system is still unclear
- It is also unclear as how the prediction system interacts with other cognitive processes/resources (e.g. working memory)

Broad goal of this project

- What are the processes underlying moment-by-moment predictions of a Hindi native speaker?
- How do these interact with working memory constraints?
- What does it tell us about human sentence processing in general?

More specifically

- We investigate:
 - verb class,
 - word order,
 - verb morphology

- In order to investigate these, we vary:
 - the number of (animate) noun phrases,
 - the case markers on the noun phrases

Experiment

- We conducted a sentence completion study
 - Items were presented using Linger in self-paced reading paradigm
 - Participants read incomplete sentences and had to complete it meaningfully
- Items were randomized
- 30 Hindi native speakers from Jawaharlal Nehru University and IIT Delhi participated in the experiment

Design

- No. of noun phrases: 1, 2, 3
 - *ladkaa, dost, mahilaa*
- Case markers: *0, ne, ko, se*
- There were:
 - 4 one noun phrase conditions
 - 16 two noun phrase conditions
 - 64 three noun phrase conditions
- There are 29 subjects, and 84 items
 - In all there were 2436 data points
 - Data from 1 subject was removed because he/she did not understand the task
 - 1 condition only had 6 data points due to a coding error

Design

- 1 NP examples
 - *ladke ne*
 - *ladke ko*
 - ...
- 2 NP examples
 - *ladke ne dost ko*
 - *ladke ko dost se*
 - ...
- 3 NP example
 - *ladke ne dost ko mahilaa se*
 - ...
- So, all items were of the form
 - *ladka/ladke.ne/ko/se dost.0/ne/ko/se mahilaa.0/ne/ko/se*

Coding

- After running the experiment, data transcription involved coding the following information based on participant response
 - **Grammaticality:** 1 (grammatical), 0 (ungrammatical)
 - **Word Order:** 1 (canonical), 0 (non-canonical)
 - **Verb class:** T (transitive), IN (intransitive), DT (ditransitive), EXP (experiencer), COP (copula), CAUS (Causative)
 - **Verb morphology:** Perf (perfective), Obl (obligational), ...
 - ...
 - ...
 - ... (some others that are not discussed here)

Caveats

- All sentences presented in isolation
 - It is known that discourse will have some role to play in predicting word order
- Broad verb classes
 - Some distinctions currently not made
- Only animate nouns used
- Current analysis of verb class/morphology only considers simple completions
 - e.g. completions with non-finite verbs are deemed complex and not considered
- Passives, argument ellipsis, imperatives, noun compounding not considered
- **These issues will be addressed in the future analysis/investigations**

Grammaticality

- If prediction is pervasive during processing, then we expect that the noun phrases and case-markers will be successfully employed to produce grammatical continuations
- So, if a condition that can otherwise be completed grammatically gets high ungrammatical continuations, we need to understand why that is

Grammaticality

- Very few ungrammatical cases in 1 NP conditions. Not surprising!
- Examples where ~50% instances are ungrammatical

2 NP	3 NP
<i>ne, ne</i>	<i>ko, se, se</i>
<i>ko, ko</i>	<i>se, ne, se</i>
<i>se, se</i>	<i>se, ko, ko</i>
	<i>se, ko, se</i>
	<i>se, se, ko</i>

Grammaticality

- Other 3 NP cases such as (*ne, ko, se*), (*se, ko, ne*), etc. are grammatically completed
- Current pattern suggests that the main reason for ungrammaticality (in cases where grammatical continuations are possible) is some kind of cue-based interference. The interference happens because of the presence of duplicate case markings, e.g. (*se, ne, se*), (*se, ko, ko*), etc.
- **If this is true, then it would mean that certain valid case-marker combinations are not always beneficial for predicting the upcoming verb**

Word Order

- Hindi is known to be a relatively free word order language
- Previous work on word order variation (e.g. Gambhir, 1981; Butt and King, 1996; Kidwai, 2000) are motivated by information structure and discourse considerations
- In this study, we only look at sentences in isolation

Word Order

- In this study, 3.5% of the data has continuations with non-canonical word order
- What are these conditions?

<i>ko, 0</i>	<i>se, ko</i>
<i>ko, 0, 0</i>	<i>se, 0, ko</i>
<i>ko, se, 0</i>	<i>se, ko, 0</i>
<i>se, 0</i>	<i>se, 0, 0</i>

- These conditions have NP1 with a *ko* or *se* case marker
- **This means that *ko/se* is not preferred in the canonical subject position**
 - Although they are grammatically permitted at this position

Word Order

- Interestingly, the order of case markers interact with word order prediction
 - *se* before *ko* in a 2 NP condition was not treated as a subject, rather *ko* was treated as the subject

Verb class

- We know that case markers and no. of NPs together determine the verb class
- Not surprisingly we found evidence for this in our data

	Dominant verb class
1 NP	Intransitive, Experiencer, Copula
2 NP	Transitive
3 NP	Causative

Verb class

- Similarly, case markers determine the verb class

	0	ne
1 NP		

- *ne* case marker insures the prediction of a transitive verb in close to 90% instances
- *ko* resulted in prediction of experiencer verbs (or verbs with dative subjects)
 - *ladke ko pyaar hai, ladke ko bukhar hai*
- *se* does not seem to affect verb class prediction as it is rarely treated as a subject

Dynamic nature of verb class prediction

- How does the verb class prediction change as we get additional NPs?

NP1.0 NP2.C*: Intransitive → Transitive

NP1.0 NP2.C* NP3.se/ko: Intransitive → Transitive → Causative

NP1.ko NP2.se: Transitive → Experiencer

NP1.ko NP2.se N3.ne: Transitive → Experiencer → Causative

NP1.ko NP2.se N3.0: Transitive → Experiencer → Transitive

NP1.se NP2.se/ko: Intransitive → Causative/Experiencer

NP1.se NP2.ko NP3.ne: Intransitive → Causative/Experiencer → Causative

Dynamic nature of verb class prediction

- We can see that Hindi native speakers are constantly revising their predictions based on the new material
- We still need to investigate how this change interacts with word order
- Studying these changes will help us determine the processing cost of these factors through an online experiment

Morphology

- We coded perfective, habitual, continuous, obligational, abilitative verb comple
- Perfective~78%
- Habitual~8%
- Continuous~3%
- Obligational+Abilitative~8%
- Data shows default completion is perfective

Morphology

- Not surprisingly, *ne* triggers perfective morphology
- *se* marker resulted in the prediction of verbs with a specific (abilitative) verb complex
 - *ladke se padha nahi gya., ladke se kaam nahi hua*
- conditions where we got habitual morphology (R data, subset HAB)
- conditions where we got continuous morphology (subset conditions for CONT)

- Data shows default completion is perfective

Wrapping up

- Our analysis reiterates many established facts wrt the effect of no. of NPs and case marker on verb class and morphology
- However, we noted that the role of these factors is constrained, i.e. case markers need not always help in prediction
- In the absence of any prior discourse, non-canonical word order is rare. *se/ko* is not treated as a subject when it occurs in a sentence initial position
- Verb class prediction can change with additional material. These changes should lead to differing processing cost for different conditions

Thanks!

