

JU-PTBSGRE: GRE Using Prefix Tree Based Structure

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Abstract

This paper presents a Prefix Tree based model of Generation of Referring Expression (RE). Our algorithm PTBSGRE works in two phases. First, an encoded prefix tree is constructed describing the domain structure. Subsequently, RE is generated using that structure. We evaluated our system using Dice, MASI, Accuracy, Minimality and Uniqueness scoring method using standard TEVAL tool and the result is encouraging.

1 Introduction

Generation of referring expression (GRE) is an important task in the field of Natural Language Generation (NLG) systems. The existing algorithms in GRE lie in two extremities. Incremental Algorithm is simple and speedy but less expressive in nature whereas others are complex and exhaustive but more expressive in nature. We propose a new Prefix Tree (Trie) based framework for modeling GRE problems. It incorporates intricate features of GRE (like set and boolean descriptions, context sensitivity, relational description etc.) while achieving attractive properties of Incremental algorithm (simplicity, speed etc.). The prefix tree based algorithm works in two phases. First, it encodes the description, stored in the knowledge base, in the form of prefix tree structure. Secondly, it generates the referring expression identifying the target object, which is basically a node search problem in the tree. The edges in our encoded trie structure are labeled and the path from root to that node forms the distinguishing description for the target object.

Let \mathbf{D} be the Domain, \mathbf{r} be the target object and \mathbf{P} be the 'PreferredAttributes' List. The Trie construction algorithm **ConstructTrie(D,P,T)** is shown in figure 1, Referring expression generation

algorithm **MakeRefExpr(r,p,T,L)** is shown in figure 2, where \mathbf{T} is a node pointer and \mathbf{p} is pointer to parent of that node. Our algorithm **MakeRefExpr** returns set of attribute-values \mathbf{L} to identify \mathbf{r} in the domain. $[[\mathbf{N}_i]] = \{\mathbf{d} \mid \mathbf{d} \in \mathbf{D} \text{ and } \mathbf{d} \text{ is stored at node } \mathbf{N}_i \text{ where } \mathbf{N}_i \text{ is an } i\text{-th level node}\}$. **Card(N)** is cardinality of set of objects in node \mathbf{N} .

```

ConstructTrie(D, P, T) {
  If (D = ∅ ∨ P = ∅)
  Then Stop
  Else
    Create a node N at T
    Set [[ N ]] = D
    Extract front attribute Ai from list P
    P' = P - { Ai }
    For each value Vj of attribute Ai do
      Create Edge Ej with label Vj as T→Nextj
      Dj = D ∩ [[ Val(Ej) ]]
      ConstructTrie(Dj, P', T→Nextj)
    End For
  End If
}

```

Figure 1. Prefix Tree Generation Algorithm

```

MakeRefExpr(r, P, T, L) {
  If (r ∉ [[ T→N ]])
  Then L ← ∅
  Return L
  Else If ({r} = [[ T→N ]])
  L = L ∪ Val(P→Ej)
  Return L
  Else If (isLeaf(T) ∧ {r} ⊂ [[ N ]])
  Then L ← ∅
  Return L
  Else {
    If (Card(P→N) > Card(T→N))
    Then L = L ∪ Val(P→Ej)
    P = T
    For each outgoing edge T→Nextj (Ej) do
      L' = MakeRefExpr(r, P, T→Childj, L)
      If (L' ≠ ∅)
      Then Return L'
    }
  }
}

```

Figure 2. Expression Generation Algorithm

The significant achievement is that incompleteness of previous algorithms can be tackled in this model in a straightforward way. For example, in case of vague descriptions (overlapping properties), Incremental and other algorithms are unable to find unambiguous description even if it exists but our prefix tree model takes into account hearer model

and generate description for identifying the target object. Besides, in case of Boolean, plural, context sensitive and relational description generation our model provides a simple and linguistically rich approach to GRE.

2 Evaluation Results

In Table 1 and 2 the evaluation results for Furniture and People data has been shown.

TRIAL	DICE	MASI	A A CCU RAC Y	UNI QUE NES S	MIN IMA LIT Y
s101t4	1.0	1.0	1.0	1.0	0.0
s101t6	0.75	0.2	0.0	1.0	0.0
s101t7	1.0	1.0	1.0	1.0	0.0
s108t5	0.85714	0.5	0.0	1.0	0.0
s114t1	1.0	1.0	1.0	1.0	0.0
s114t6	1.0	1.0	1.0	1.0	0.0
s120t4	0.57142	0.13333	0.0	1.0	0.0
s120t6	0.57142	0.13333	0.0	1.0	0.0
s123t1	0.0	0.0	0.0	1.0	0.0
s131t5	1.0	1.0	1.0	1.0	0.0
s169t1	0.8	0.44444	0.0	1.0	0.0
s169t3	1.0	1.0	1.0	1.0	0.0
s18t1	1.0	1.0	1.0	1.0	0.0
s191t3	0.66666	0.16667	0.0	1.0	0.0
s191t4	1.0	1.0	1.0	1.0	0.0
s191t6	0.57142	0.13333	0.0	1.0	0.0
s195t1	0.4	0.08333	0.0	1.0	0.0
s197t1	1.0	1.0	1.0	1.0	0.0
s197t3	0.85714	0.5	0.0	1.0	0.0
s197t4	1.0	1.0	1.0	1.0	0.0
s197t6	0.75	0.2	0.0	1.0	0.0
s19t5	0.57142	0.13333	0.0	1.0	0.0
s19t7	0.5	0.11111	0.0	1.0	0.0
s20t5	1.0	1.0	1.0	1.0	0.0
s20t6	0.66666	0.33333	0.0	1.0	0.0
s22t3	0.85714	0.5	0.0	1.0	0.0
s22t4	0.85714	0.5	0.0	1.0	0.0
s28t6	0.57142	0.13333	0.0	1.0	0.0
s307t3	0.8	0.44444	0.0	1.0	0.0
s30t4	1.0	1.0	1.0	1.0	0.0
s310t3	0.57142	0.13333	0.0	1.0	0.0
s312t5	1.0	1.0	1.0	1.0	0.0
s313t2	1.0	1.0	1.0	1.0	0.0
s315t2	0.8	0.44444	0.0	1.0	0.0
s31t4	0.33333	0.06667	0.0	1.0	0.0
s31t6	0.28571	0.05555	0.0	1.0	0.0
s320t2	0.66666	0.16667	0.0	1.0	0.0
s320t4	1.0	1.0	1.0	1.0	0.0
s320t7	0.57142	0.13333	0.0	1.0	0.0
s323t1	1.0	1.0	1.0	1.0	0.0
s323t3	1.0	1.0	1.0	1.0	0.0
s326t2	0.5	0.22222	0.0	1.0	0.0
s326t6	1.0	1.0	1.0	1.0	0.0
s329t5	0.8	0.44444	0.0	1.0	0.0
s331t6	0.57142	0.13333	0.0	1.0	0.0
s346t2	0.33333	0.06667	0.0	1.0	0.0
s367t4	1.0	1.0	1.0	1.0	0.0
s367t5	1.0	1.0	1.0	1.0	0.0
s373t5	1.0	1.0	1.0	1.0	0.0
s373t6	0.66666	0.33333	0.0	1.0	0.0
s374t4	1.0	1.0	1.0	1.0	0.0
s374t5	0.57142	0.13333	0.0	1.0	0.0
s41t2	1.0	1.0	1.0	1.0	0.0
s41t5	1.0	1.0	1.0	1.0	0.0
s48t4	0.66666	0.16667	0.0	1.0	0.0
s49t4	1.0	1.0	1.0	1.0	0.0
s49t6	0.5	0.11111	0.0	1.0	0.0
s50t6	0.85714	0.5	0.0	1.0	0.0
s53t1	1.0	1.0	1.0	1.0	0.0
s53t2	1.0	1.0	1.0	1.0	0.0
s53t4	1.0	1.0	1.0	1.0	0.0
s57t5	0.57142	0.13333	0.0	1.0	0.0
s57t6	0.66666	0.16667	0.0	1.0	0.0
s59t3	1.0	1.0	1.0	1.0	0.0
s60t2	0.8	0.44444	0.0	1.0	0.0
s62t1	1.0	1.0	1.0	1.0	0.0
s62t5	0.33333	0.06667	0.0	1.0	0.0
s62t6	0.5	0.11111	0.0	1.0	0.0
s73t1	1.0	1.0	1.0	1.0	0.0
s73t3	1.0	1.0	1.0	1.0	0.0
s73t6	1.0	1.0	1.0	1.0	0.0
s73t7	0.57142	0.13333	0.0	1.0	0.0
s78t4	0.66666	0.33333	0.0	1.0	0.0
s78t7	0.57142	0.13333	0.0	1.0	0.0
s81t2	1.0	1.0	1.0	1.0	0.0
s81t4	1.0	1.0	1.0	1.0	0.0
s84t1	1.0	1.0	1.0	1.0	0.0
s84t6	0.66666	0.33333	0.0	1.0	0.0
s98t5	1.0	1.0	1.0	1.0	0.0
s98t7	1.0	1.0	1.0	1.0	0.0

Table1: Evaluation Result of Furniture data

TRIAL	DICE	MASI	A A CCU RAC Y	UNI QUE NES S	MIN IMA LIT Y
s101t24	1.0	1.0	1.0	1.0	0.0
s102t23	1.0	1.0	1.0	1.0	0.0
s106t23	0.0	0.0	0.0	1.0	0.0
s114t21	0.85714	0.5	0.0	1.0	0.0
s114t23	0.66667	0.33333	0.0	1.0	0.0
s123t25	0.28577	0.05556	0.0	1.0	0.0
s131t21	1.0	1.0	1.0	1.0	0.0
s131t22	1.0	1.0	1.0	1.0	0.0
s132t24	0.4	0.08333	0.0	1.0	0.0
s132t25	0.28571	0.05556	0.0	1.0	0.0
s169t26	0.85714	0.5	0.0	1.0	0.0
s177t22	0.8	0.44444	0.0	1.0	0.0
s181t24	1.0	1.0	1.0	1.0	0.0
s18t25	0.85714	0.5	0.0	1.0	0.0
s191t22	0.4	0.08333	0.0	1.0	0.0
s191t24	1.0	1.0	1.0	1.0	0.0
s192t25	0.66667	0.16667	0.0	1.0	0.0
s197t22	0.85714	0.5	0.0	1.0	0.0
s197t23	0.33333	0.06667	0.0	1.0	0.0
s197t26	0.33333	0.06667	0.0	1.0	0.0
s19t25	0.28571	0.05556	0.0	1.0	0.0
s20t22	0.8	0.44444	0.0	1.0	0.0
s20t26	0.85714	0.5	0.0	1.0	0.0
s22t25	0.85714	0.5	0.0	1.0	0.0
s307t22	0.66667	0.16667	0.0	1.0	0.0
s30t25	1.0	1.0	1.0	1.0	0.0
s310t25	0.57143	0.13333	0.0	1.0	0.0
s312t25	0.85714	0.5	0.0	1.0	0.0
s315t21	0.66667	0.33333	0.0	1.0	0.0
s315t24	1.0	1.0	1.0	1.0	0.0
s31t24	0.4	0.08333	0.0	1.0	0.0
s31t26	0.33333	0.06667	0.0	1.0	0.0
s320t23	1.0	1.0	1.0	1.0	0.0
s323t23	1.0	1.0	1.0	1.0	0.0
s323t26	0.66667	0.16667	0.0	1.0	0.0
s331t23	1.0	1.0	1.0	1.0	0.0
s332t26	0.85714	0.5	0.0	1.0	0.0
s373t23	1.0	1.0	1.0	1.0	0.0
s373t26	0.85714	0.5	0.0	1.0	0.0
s374t21	1.0	1.0	1.0	1.0	0.0
s374t22	0.85714	0.5	0.0	1.0	0.0
s374t23	0.66667	0.33333	0.0	1.0	0.0
s40t26	0.4	0.08333	0.0	1.0	0.0
s41t23	0.4	0.08333	0.0	1.0	0.0
s48t21	0.75	0.2	0.0	1.0	0.0
s49t21	0.28571	0.05556	0.0	1.0	0.0
s49t22	0.4	0.08333	0.0	1.0	0.0
s49t23	0.8	0.44444	0.0	1.0	0.0
s50t24	1.0	1.0	1.0	1.0	0.0
s50t25	0.85714	0.5	0.0	1.0	0.0
s50t26	0.85714	0.5	0.0	1.0	0.0
s53t21	1.0	1.0	1.0	1.0	0.0
s53t23	0.66667	0.33333	0.0	1.0	0.0
s53t26	0.85714	0.5	0.0	1.0	0.0
s57t23	0.4	0.08333	0.0	1.0	0.0
s57t24	0.4	0.08333	0.0	1.0	0.0
s57t25	0.28571	0.05556	0.0	1.0	0.0
s59t21	0.66667	0.33333	0.0	1.0	0.0
s59t24	1.0	1.0	1.0	1.0	0.0
s59t25	0.85714	0.5	0.0	1.0	0.0
s60t24	0.57143	0.26667	0.0	1.0	0.0
s62t21	0.0	0.0	0.0	1.0	0.0
s69t25	0.0	0.0	0.0	1.0	0.0
s73t25	0.85714	0.5	0.0	1.0	0.0
s78t21	0.85714	0.5	0.0	1.0	0.0
s78t25	0.88889	0.53333	0.0	1.0	0.0
s81t24	1.0	1.0	1.0	1.0	0.0
s81t25	0.85714	0.5	0.0	1.0	0.0

Table2: Evaluation Result of People data

References

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