US5560 Knowledge Discovery and Management

Problem Set (PS-1A) June 5, 2017

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We are supposed to build a knowledge graph for the following input (text data).

1. Describe your knowledge about knowledge graph.

Why do we want to build such a knowledge graph?
 What steps are required? Show your own workflow for this task.

4. What are the challenges?

5. Draw a knowledge graph for the given data.

CHICAGO (AP) — Citing high fuel prices, United Airlines said Friday it has increased fares by \$6 per round trip on flights to some cities also served by lower-cost carriers. American Airlines, a unit AMR, immediately matched the move, spokesman Tim Wagner said. United, a unit of UAL, said the increase took effect Thursday night and applies to most routes where it competes against discount carriers, such as Chicago to Dallas and Atlanta and Denver to San Francisco, Los Angeles and New York.

hnowledge graphs are graphs of interlinked entities and their attributes. Knowledge graphs have an ontology as its schema.

(2) a) The knowledge graph are made of interlinked entities. It provides a straightforward applicach where users an understand these entities by directly providing can understand these entities by directly providing key facts. (For example, information about Kansas City)

b) For questions for which answers are not directly available, the users call use efficient exploitation techniques and locate the right portion of the knowledge graph to get the answer.

(For example, what were the 5 coldest years in Kansas (ity in the last 15 years?)

c) Linking multiple knowledge graphs might help in decision making in large organization.

So by building a broundge graph.

-> We are improving the learning curve.

-> graph is the most yenible data structure. So, quesing and information setrieval will be easy.

- minimizes redundany.

- improves decision-making.

3 Step: 1 Natural language Processing.

Annotation - lemmatization -> POS -> Named Estity Recognition.

CHICAGO -> CHICAGO -> NNP -> LOCATION

-LRB- -> -LRB--ARP -> NER-> ORGANIZATION. -RRB- -> -83b- -> - RRB-

Citing -> like -> VBS high -> high -> JJ hul -> hule -> NN

gred

pice -> NNS pièces -> With > wited > NNP -> ORGANIZATION

--> NNPS -> ORGANIZATION Annotation -> bennatization > 105 -> Named Editity > NNP -> DATE > MONGY > MONEY V8D NNS increase -> VBN > 182 fights - fight -> NNS sound -> NN thip IN 38WE -> VBD SI T KO Some -> DT > cty > NNS Sistines also - RB N/ A B tos + Friday have fare ha Z 1 to Friday incleased Said Joses ng They seaved has Sound Sa some 18 Whiles also Bu

- United -> NNP -> ORGANIZATION American -> American -> NNP. -> ORGANIZATION dillines -> dislines -> NNPS -> ORGANIZATION Wagnes -> Nagnes -> NNP -> PERSON Said -> Soup -> VBD Tim -> PERSON downs-cost - Jours-cost - 5J inmediately -> inmediately -> RB cassies -> cassies -> NNS spokesmen - spokesmen >NN matched > match > VBD

the > the > DT

move > move > NN a d d d DT

with d with d NN

AMR d D D. wit - with - NN 1 -United

Thursday -> NNP -> DATE night -> NN -> TIME apply > VBZ meet -> JJS -> soute -> NNS exect JNN such -> such -> JJ take >180 cassiess - cassies -> NNS where -> NRB increase J NN competes -> compete -> VBZ it -> PRP against - against -> IN discount -> discount -> NN and 108 V the Thursday -> month 1 1 increase > and the effect applies Soid where 如 most 26 soutes t,

- LOCATION

chicago -> Chicago -> IN

Dallas -> Dallas -> NNP-> LOCATION Les - LOCATION and -> and -> CC otherta -> otherta -> NNP -> LOCATION Denver -> Denver -> NNP -> LOCATION -> Angeles -> NNP -> LOCATION San - San - NNP -> LOCATION Francisco -> Francisco -> NNP -> LOCATION -> NEW -> NNP -> LOCATION and -> and -> CC to -> to -> 70 -> and -> Cc Angeles and

ability of systems to plocess sentences in a natural Janguage such as English, hather than in a specialized astificial computer Janguage such as C++. 1 Tokuization : tokenzing woods that into words. Natural danguage processing refers to the use and It involves the following stops:

@ fermatization/Berning: St

hemmetizedien is the peaces of grouping together. The different injected forms of a word so they can be sterning is the peaces of reducing words to their sten, base or root form- generally a written word analysed as a single term.

c) pos Togging / Chunking

Toget per this powers. Tagging. Used a pee-defined

It is the process of analyzing a sentence by taking each word and determining its structure from its constituent parts. It uses two components: a peases and a gammas.

e) Name Estily Recognition.

Recognition Information extraction that seeks to locate and classify named entities in tent into pee-defined extegories such as , monetary names of pessons, organizations, locations It is a subtask of Norrad-Edityenpessions of times, quantities Volues, percentages etc

expressions that sofe to the same entity in a text. It is important stop for higher NeP tasks like document summarization, question It is the task of finding all arsweing and information extraction. 4) (0- seference Resolution

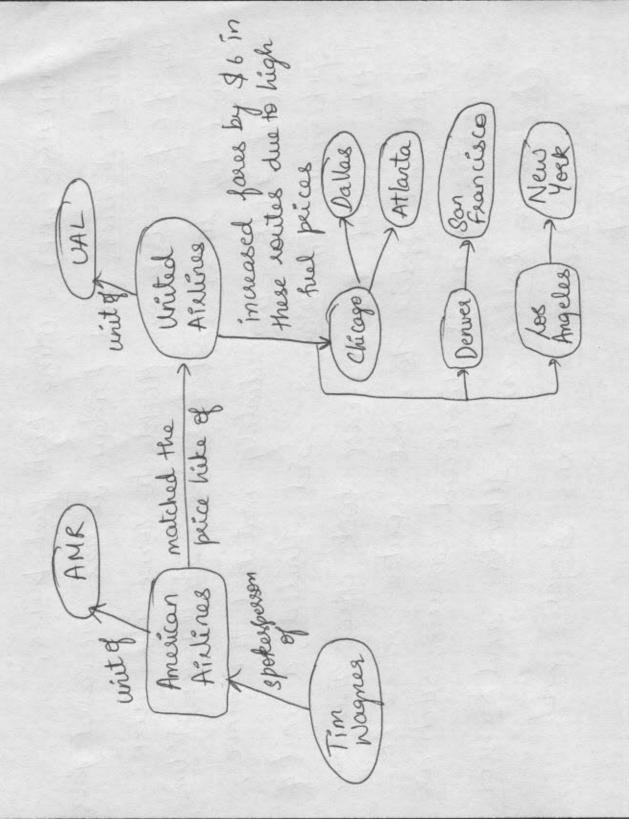
Step 2: Information Retrieval and Intraction.

United Milines said Fieday it has increased fores bewed by lower-cost cassiers. American stillings Chicago to Ballas and Atlanta and Denner spokesman. Tim Wagnes said. United, a unit of UAL, said the increase took effect Thussday competes against discount cassiers, such as a unit "AMR, immediately not thed the move, wight and applies to most loutes where it by \$ 6 per trip on Mights to some cities also to San Francisco, dos Angeles and CHICAGO (AP) - ching high fuel pieces, New York

step 3: Topic discovery

Places	Chicago Ballas A-Hanta Berves San Francisco Les Angelas yerk
People	Jagner.
Reganization	United distinct Tim I American distincts ANR UAL

Knowledge (Raph Construction



Correctness of the knowledge graph. The accusate Completeness of the knowledge graph. These can be a lot of missing information. The knowledge graphs from different sources will adhere to different formats. generate these knowledge graphs. As a result These is no universal algorithm to 4) Challenges of Knowledge Graph. data.

Public knowledge bases may not have the secusate deta.

3 plans knowledge graph

P.T.0