



NII-UIT at TRECVID 2023: Deep video understanding

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Overview

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DVU Task Introduction

Input

- Full length feature movie.
- List of scenes (segmented manually based on semantic)
- Ontology (vocabulary) (for the entire dataset)
- List of entities (Person location)
- 5-7 images of each entities
- Transcript of movies (generated by <u>OpenAl's whisper</u>)

• A barrage of queries (2 types movie level, 4 types scene level)

Movie level queries

Fill in the graph query

Given a list of edges (relation) (relationship, event, action, etc...) for node (entity) X. Some (0-3) of those edges may have its target left BLANK.

Return a ranked list of of candidate for node X

Movie level queries

Multiple choice Q&A (151 of them)

- One natural language question
- 6 natural language choices per question
- More questions and higher difficulty than 2022

Type 1: Find unique scene (optional in 2022) (40 of them)

Given a **full**, **inclusive** list of interaction between persons (unknown) in a scene, find that scene number.

Type 2: Find person (not required) 18 of them

Given scene number, list of interactions (both to and from) a person X, find X

Scene level queries

Type 3 Find Next interaction (20 of them)

Given a reference scene number, 1 interaction between 2 People X and Y in the scene, and a target scene number.

Find the immediate next interaction between X and Y in target scene number. (Target scene and reference number can be the same

Type 4: Find previous interaction (20 of them)

Type 1,2,3,4 have to be submit in one group.

Type 5: Match scene to text (50 of them)

Given 1 natural language sentence description of a scene and 10 scene numbers. Find the number of the scene best match the description.

Type 6: Scene sentiment classification (50 of them)

Given 1 scene number, a list of 6 sentiment words. Find the correct sentiment label for that scene.

Dataset

	time_of_movi		time_of_scene (s)		
	е				
video_name	(hh:mm:ss)	# scene	min	max	avg
Memphis	1:18:39	47	17	294	97
Archipelago	1:50:04	57	21	389	113
Bonneville	1:32:39	41	19	269	124
heart_machine	1:23:37	28	22	451	158
Little_Rock	1:22:48	39	24	289	121



Challenges

- > Very difficult questions (especially this year test set)
 - Required very deep multimedia analysis of the movie to answer
 - Difficult even to human (could required a rewatch of the whole movie to answer a question)
- Full-length-feature-movie
 - Movies are long and diverse in genre
 - Required real-world knowledge to understand
 - Understanding scenes require whole movie level information

Our approach

Scene based entity recognition

- → For 'character' entities:
 - MTCNN + ArcFace
- → For 'Location' entities:
 - ResNet 50

Movie - query 1 - Find entity from relations

Only considered "share-scene" relation

Entity rarely have relations with itself

- 1. Get the list of candidate_entities, i.e not mentioned in query.
- 2. Sort the list of candidate by the number of scenes they share with any mentioned entities
- 3. Return the ranked list

Movie - query 2 - free form Q&A

Use google's pre-trained universal sentence encoder model. https://tfhub.dev/google/universal-sentence-encoder-qa/3

Run 1:

- Use the whole movie's ASR text as context.

Run 2:

 Use only the scene where mentioned entities appears as context

Scene level query type 4-5

Just use Google Universal sentence encoder to match the scenes' ASR with query

- Type 4: Match given description with each scene's ASR in the choices, return highest match
- Type 5: Match given scene's ASR with each sentiment label, return highest match.

Submission result

Movie level - query 1



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Movie - query 2 - natural language Q&A



Human generated question





Scene query4-Match scene with description



Scene query - Scene sentiment labeling





Conclusion

- A very simple baseline was proposed for DVU task.
 - Leveraging visual entity recognition and LLM for ASR processing.
 - Stronger baseline than expected
- Very challenging task