

Motivation

- Significant progress in **image-based** VQA with various datasets



How about VQA tasks on **videos**?

- A **few** datasets use **movie** as data source for video VQA

MovieQA
[Tapaswi et al., CVPR 2016]

Patrick

Q) How does Patrick start winning Kat over?
 A) By knowing Kat's likes and dislikes

LSMDC 16
[Rohrbach et al., IJCV 2017]

Richard Parker

Q) Richard Parker _____ from the boat.
 A) watches

- Solutions require information not available in visual content.
 eg) context, sound, script

Can we define **new tasks** specifically for **video VQA**?

New Tasks for Video VQA

1. Counting Repetitions



2. Reasoning State Transitions



A New Dataset for Video VQA: 165K QA Pairs

a) Repetition Count: 30K New Tasks for Video QA

Q) How many times does the man wrap string?
 A) 5 times

b) Repeating Action: 23K

Q) What does the duck do 3 times?
 A) Shake head

c) State Transition: 59K

Q) What does the bear on right do after sitting?
 A) Stand

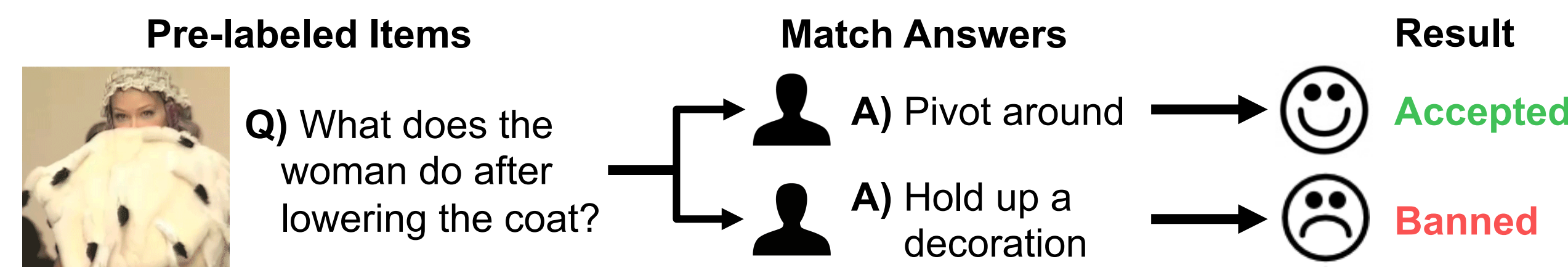
d) Frame QA: 53K

Q) What is dancing in the cup?
 A) Tree

Methods for Generating QA Pairs

a-c) Template-based. **Crowdsourced** via **amazon mechanical turk**

- Strict quality control: Blacklist workers based on pre-labeled items

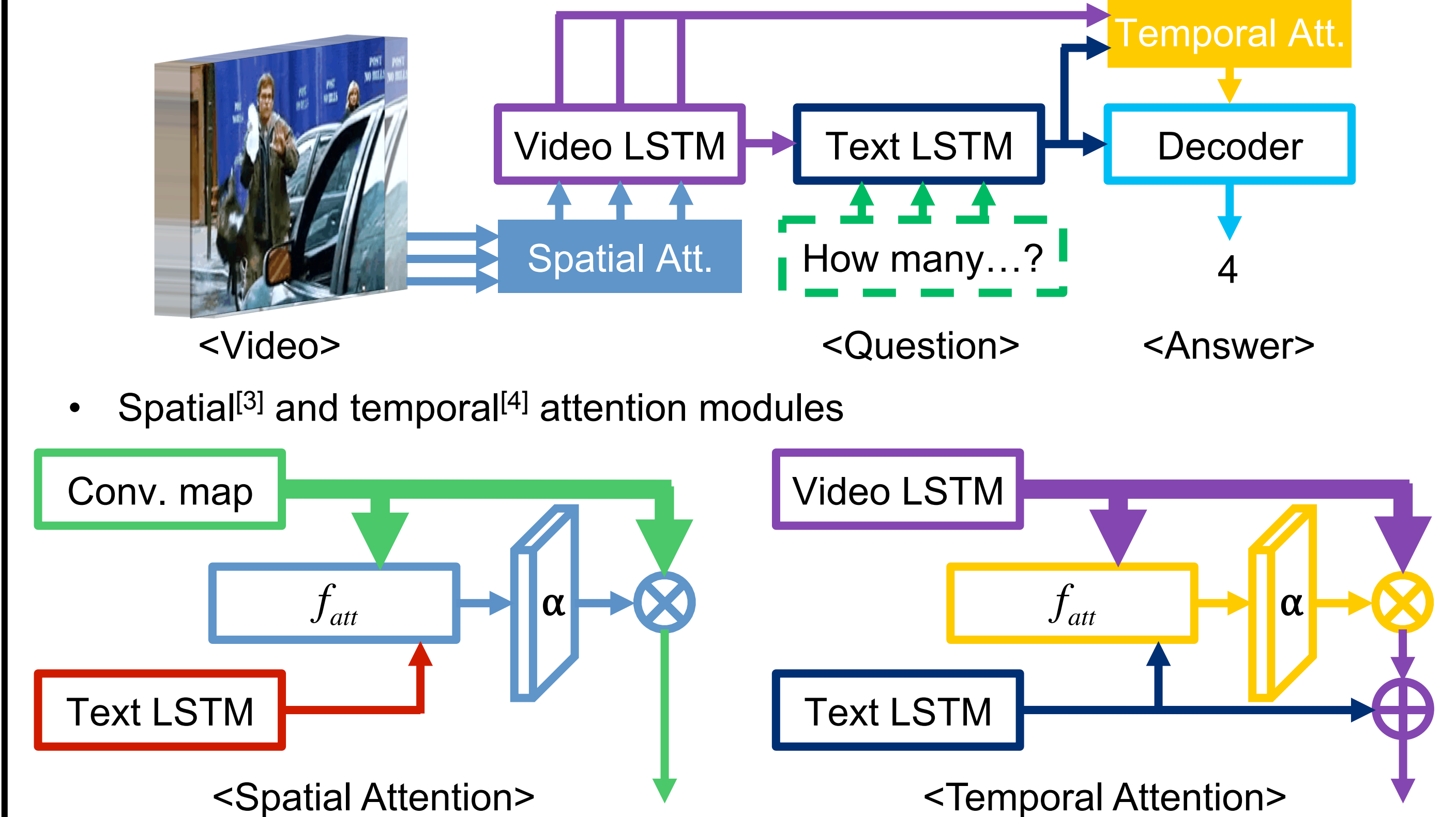


- Synonyms are considered as correct answers
- Generate four wrong answers based on a cosine similarity of the verbs

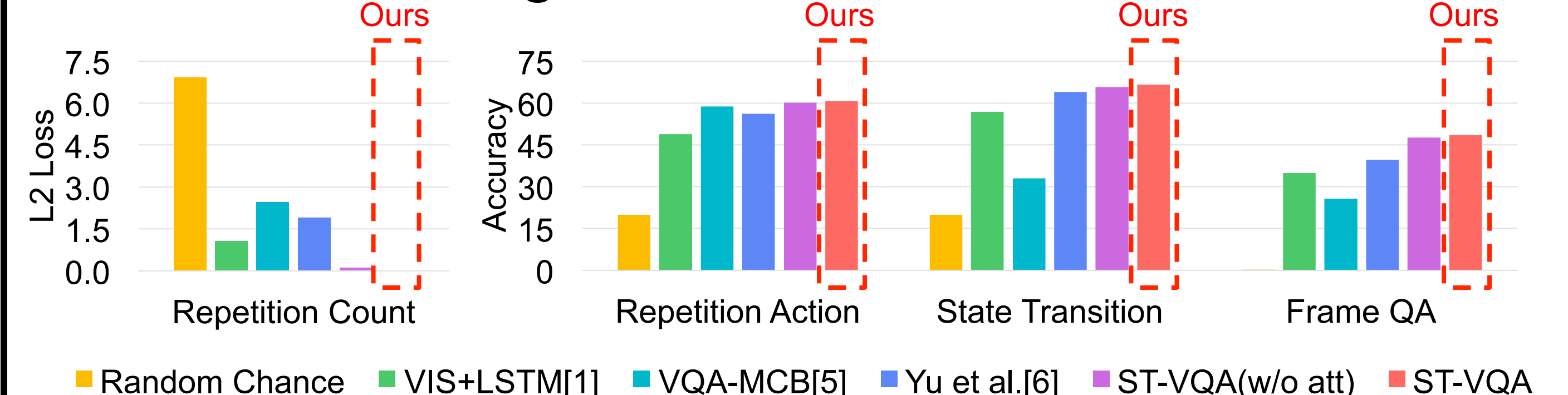
d) NLP-based QA generation^[1] using descriptions from TGIF dataset^[2]

- Convert a declarative sentence to an interrogative sentence

A Novel Model for Video VQA: ST-VQA



Results and Findings



- Video-based model** works better than image-based models
- Our model with **ST-attention** shows the best result.

Solving our tasks require **spatio-temporal reasoning**

References

- Ren et al., *Exploring Models and Data for Image Question Answering*, in NIPS 2015
- Li et al., *TGIF: A New Dataset and Benchmark on Animated GIF Description*, in CVPR 2016
- Xu et al., *Show, Attend and Tell: Neural Image Caption Generation with Visual Attention*, in ICML 2015
- Bahdanau et al., *Neural Machine Translation by Jointly Learning to Align and Translate*, in ICLR 2015
- Fukui et al., *Multimodal Compact Bilinear Pooling for Visual Question Answering and Visual Grounding*, in EMNLP 2016
- Yu et al., *End-to-end Concept Word Detection for Video Captioning, Retrieval, and Question Answering*, in CVPR 2017