### TutoAI: a Cross-domain Framework for AI-assisted Mixed-media Tutorial Creation on Physical Tasks

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### How-to videos are popular

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How-to videos earn the **most attention** of any content category on YouTube, even more than music clips or gaming

Think with Google

Google/Ipsos, U.S., Video Mobile Diary (n of 18,219 total video occasions), 2017.

### How-to videos are popular

#### Make pancakes



Apply makeup



Repair cars



### Pain points of watching how-to videos



Lack of information overview

Timeline scrubbing is imprecise & tedious



Liu et al. ConceptScape: Collaborative concept mapping for video learning. CHI'18; Yang et al. Improving Video Interfaces by Presenting Informational Units of Videos. CHI EA'22



Truong et al. Automatic generation of two-level hierarchical tutorials from instructional makeup videos. CHI'21; Nawhal et al. VideoWhiz: Non-Linear Interactive Overviews for Recipe Videos. GI'19



Video Section



Liu et al. ConceptScape: Collaborative concept mapping fo



Control Section				
Actions				
TOTAL	0:00-8:09			
tool (stock) (onions third) (bow) (tongs) (tin foil) (tortila) (dish) (try	pan			
(cheese (blue cheese) (cheese (mozzarella) (salad) (butter) (wooden spoon	spoon			
(yogurt) (knile) (onions) (brush) (tablespoon) (cutting board) (wrap) (griddle	e) plate			
$\left( mixed spinach, rocket and pea shoots \right) \left( fork \right) \left( seasoning \right) \left( cheese \right)$				
introducing recipe	0.09-0.50			
introducing ingredients	0.55-120			
heat up pan	121-128			
seasoning steak	125-141			
roast steak	1.42-4.10			
fry onions	158-4:00			
move onions to tinfoil	4:00-4:09			
move steak to tinfoil	435-432			
put cheese on the steak	426-433			
slice steak with cheese	4.47.511			

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### Auto-generated mixed-media tutorials are of low-quality



Search

Q



tps://www.youtube.com/watch?y=9i45Hrg57Wo

### Creating high-quality tutorials require domain knowledge

For makeup tutorials, cluster steps by facial parts





#### EYES



25. I'm going to use this browsing my benefit. And I'm ...

#### LIPS



26. I'm going to take lip land cream corset by Samantha. And ...

### AI presents opportunities to create cross-domain tutorials



### Challenges for AI-assisted cross-domain tutorial creation

**Input:** video, transcript **Output:** mixed-media tutorials (videos, images, text, diagrams)

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- Vast model space: multiple models are required and applicable
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### **Research questions:**

- How to **identify, evaluate,** and **select** models to create crossdomain mixed-media tutorials?
- How to design **user interfaces** to refine AI-generated mixed-media tutorials?

Components	Steps text, video frames, timestamps	



24. I'm gonna take my favorite blush captivating by tarped.

Components	Steps text, video frames, timestamps		Objects text labels, image	×
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#### Things You'll Need





	Hose
~	Roof cement
	Chisel
	Hammer

Components	Steps text, video frames, timestamps		Objects text labels, image	X	Dependencies temporal order, spatial order	
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Examples:



1000	24. I'm gonna take my favorite blush	
100	captivating by tarped.	





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Models identify applicable models	Predict step boundaries shot boundary detection, NLVL, video summarization, LLM	Extract objects from transcript POS tagging, LLM	Match objects to steps string match
	Generate text summary extractive & abstractive summarization, dense captioning	Locate objects in frame object detectors	Build directed acyclic graph connect steps with shared objects

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### An object component example: "chicken breast"

"...here are the ingredients...this is the boneless skinless chicken breast thinly slice like this..."



https://youtu.be/ntiGX3X-spA

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"...here are the ingredients...this is the boneless skinless **chicken breast** thinly slice like this..."



## How to extract "chicken breast" from text and images?

https://youtu.be/ntiGX3X-spA

### Candidate ML pipelines for object extraction



### Pipeline #1: general object detectors



### Candidate ML pipelines for object extraction



### Pipeline #2: part-of-speech tagging

input: "...here are the ingredients...this is the boneless skinless chicken breast thinly slice like this..."

Output (only keep nouns): "ingredients", "chicken", "breast"

https://github.com/flairNLP/flair

### Candidate ML pipelines for object extraction



### Pipeline #3: large language model (LLM) prompting

**Input:** the entire transcript

**Prompt:** "Identify the ingredients in this tutorial"

**Output:** "Boneless skinless chicken breast", ...

### Pipeline #3: detect objects given text labels



Radford et al. "Learning transferable visual models from natural language supervision." ICML 2021

### Pipeline #3: open-vocabulary object detection



#### Labels extracted by LLM:

- Boneless skinless chicken breast
- Chopped green pepper
- Roasted peanuts
- Red chili
- Green onions

Li et al. "Grounded language-image pre-training." CVPR 2022

### Candidate ML pipelines for step extraction

Step extraction



### Evaluation: a large-scale cooking video dataset

Dataset:



20 diverse instructional videos (crafting, makeup, repair)
# Evaluation: a large-scale cooking video dataset

Dataset:

□ 347 annotated cooking videos



□ 20 diverse instructional videos (crafting, makeup, repair)

Metrics:

- □ Objective:
  - □ F1 scores
  - □ ROUGE scores
  - □ Temporal Intersection over Union (tIOU)
- □ Subjective:
  - □ correction efforts required from humans

#### IDENTIFY STEPS > CHOOSE THUMBNAILS > SELECT OBJECTS > CROP OBJECTS > BUILD DEPENDENCIES

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00:01 - 00:19	Introduction to the project and purpose of creating a bumblebee seesaw	8
00:20 - 01:22	Gathering materials including a used tire and plywood board	
01:33 - 02:54	Measuring and marking the board for the curved seating areas	
02:55 - 03:38	Cutting and sanding the board	] 🚫
03:39 - 06:31	Painting the board with black and yellow stripes using painter's tape	
06:32 - 08:02	Cutting the tire in half and priming and painting it with black and yellow stripes	<b></b>
08:03 - 09:06	Creating wood blocks to attach the tire to the board	
09:07 - 10:46	Attaching the tire and handles to the board	] 🚫
	 <b>-</b>	۹ (

Instructions

#### TutoAI vs. YouTube Chapters

#### **TutoAI** components received higher ratings than **YouTube** Chapters (**N=12**)



Component quality **before editing** 

#### Component usefulness after editing

#### YouTuber's testimonial

"I think this is a great tool. I like this a lot...this is giving me the ability to do a lot more, especially creating the **flow charts** ...viewers would get a lot out of this as opposed to just a regular chapter..."



#### SevenFortyOne Radios and Repairs

SevenFortyOne · 43.1K subscribers · 590 videor

Iam Radio demonstration and repair with some Chevy Suburban and home repairs mixed i... 🔉

witter.com/SevenFortyOneB and 1 more link



Adopt a multi-modal perspective

INITIALLY ————

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Simplify complex creation by guiding and constraining user actions

Separate content from styles

8 DURING INTERACTION ———

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Focus on user-centric model selection

Support graceful degradation

▲ WHEN WRONG -

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8 DURING INTERACTION ————

Leverage strong models for cross-modal enhancement

© OVER TIME ------

### Limitations and future work



#### Object

Chang et al. "Rubyslippers: Supporting content-based voice navigation for how-to videos." CHI 2021.

### Limitations and future work





Chang et al. "Rubyslippers: Supporting content-based voice navigation for how-to videos." CHI 2021.

# Acknowledgment







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