# Multilingual Generative Language Models for Zero-Shot Cross-Lingual Event Argument Extraction



Kuan-Hao Huang\*<sup>1</sup>, I-Hung Hsu\*<sup>2</sup>, Premkumar Natarajan<sup>2</sup>, Kai-Wei Chang<sup>1</sup>, Nanyun Peng<sup>1,2</sup>

<sup>1</sup> University of California, Los Angeles
 <sup>2</sup> Information Science Institute, University of Southern California



## **Event Argument Extraction**

• Goal: extract the participants for an event described in a sentence

warplanes, witnesses said.

• Input: sentence, event trigger

- Output: (role, argument) pairs
- Challenge:
  - Dependency
  - Event structure

Event structure

Agent coalition

Victim civilians, woman

Instrument missile

Place houses

Trigger for a

Life:Die event

Five Iraqi civilians, including a woman, were killed Monday when their houses were hit by a missile fired by the US - led coalition

## Zero-Shot Cross-Lingual Event Argument Extraction

- Training examples come from source languages
- Testing examples come from target languages
- Challenge:
  - Discrepancy between language properties
  - Different vocabularies, different word order and grammar, etc.

civilians, woman
missile
houses

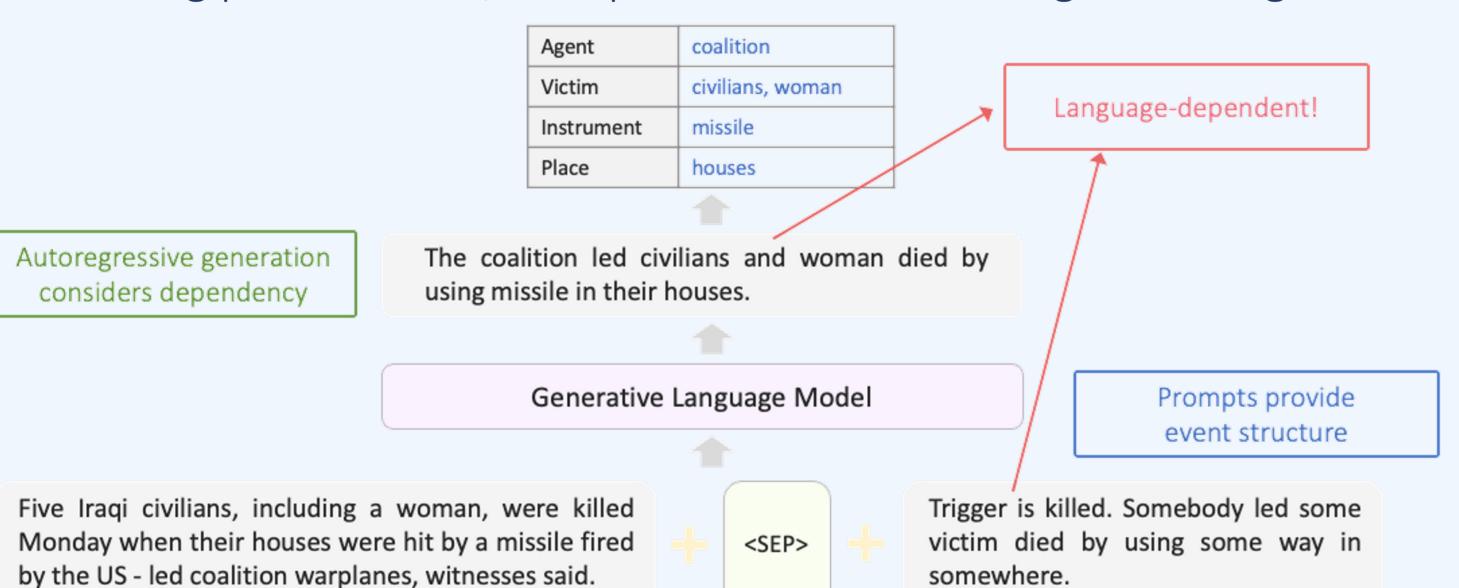
Agent 以军
Victim 青年
Instrument 催泪弹, 子弹, 实弹
Place None

Five Iraqi civilians, including a woman, were killed Monday when their houses were hit by a missile fired by the US - led coalition warplanes, witnesses said.

巴勒斯坦人持续以石块攻击以色列的部队,以军则是还以催泪弹、橡皮子弹甚至是实弹,结果又造成两名巴勒斯坦青年<mark>丧生</mark>, 10多人受伤。

#### **Previous Approaches**

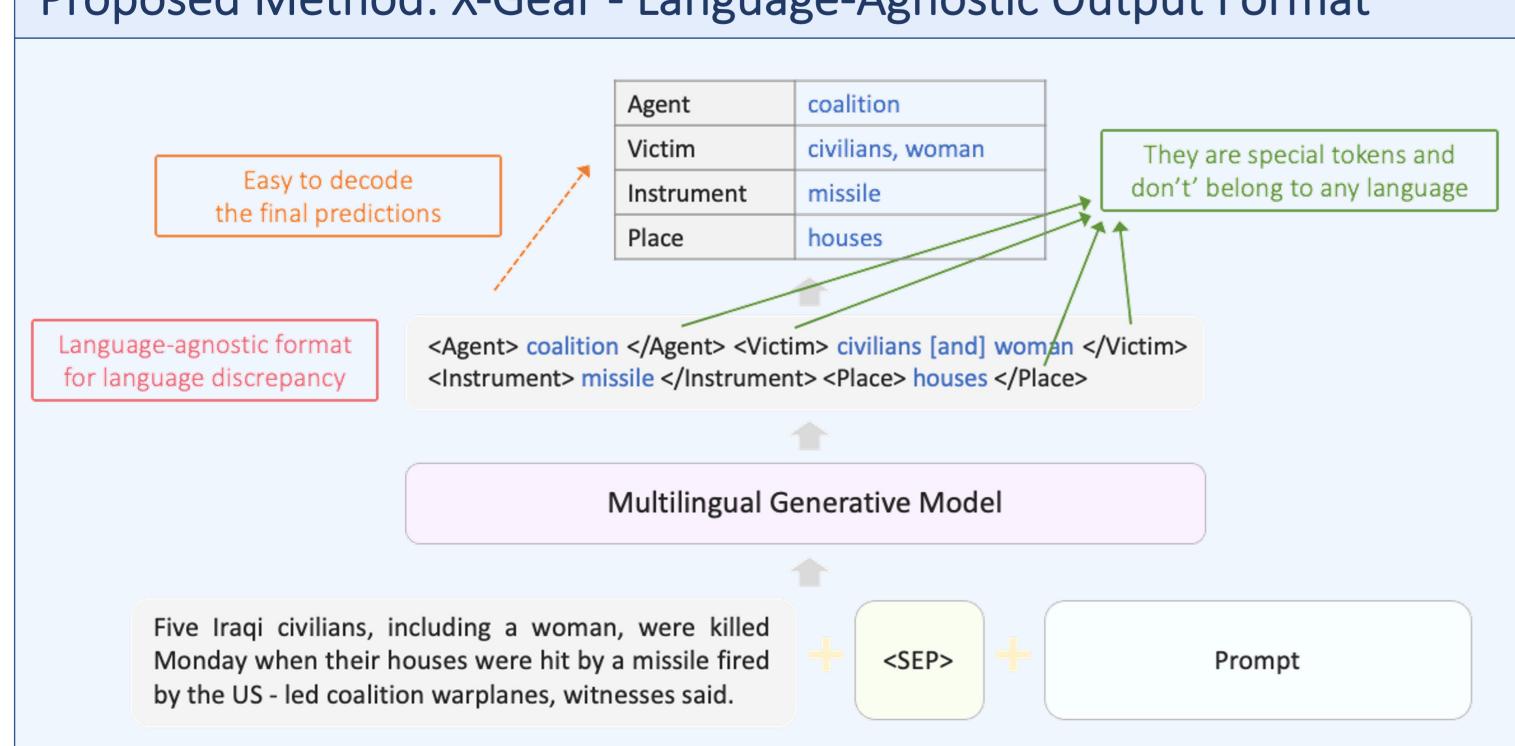
- Most previous approaches are classification-based models
  - GATE [Ahmad+ 2021] and CL-GCN [Subburathinam+ 2019]
- Recently, several works have shown that generation-based models perform better than classification-based models
  - DEGREE [Hsu+ 2022], BART-Gen [Li+ 2021], and TANL [Paolini+ 2021]
  - Strong performance, but specific to the monolingual setting



#### Our Goal

- We aim to explore the possibility of applying generation-based models for zero-shot cross-lingual event argument extraction
  - Output design
  - Prompt design

## Proposed Method: X-Gear - Language-Agnostic Output Format



#### Proposed Method: X-Gear - Prompt Design and Copy Mechanism coalition civilians, woman Victim missile Instrument Place houses <Agent> coalition </Agent> <Victim> civilians [and] woman </Victim> <Instrument> missile </Instrument> <Place> houses </Place> Provide event structure and control signal Multilingual Generative Model Five Iraqi civilians, including a woman, were killed <Trigger> killed <Template> <Agent> [None] </Agent> Monday when their youses were hit by a missile fired <SEP> <Victim> [None] </Victim> <Instrument> [None] by the US - led coalition warplanes, witnesses said. </l></l></l></l></l><

## Main Experiments

Argument Classification F1 on ACE-2005

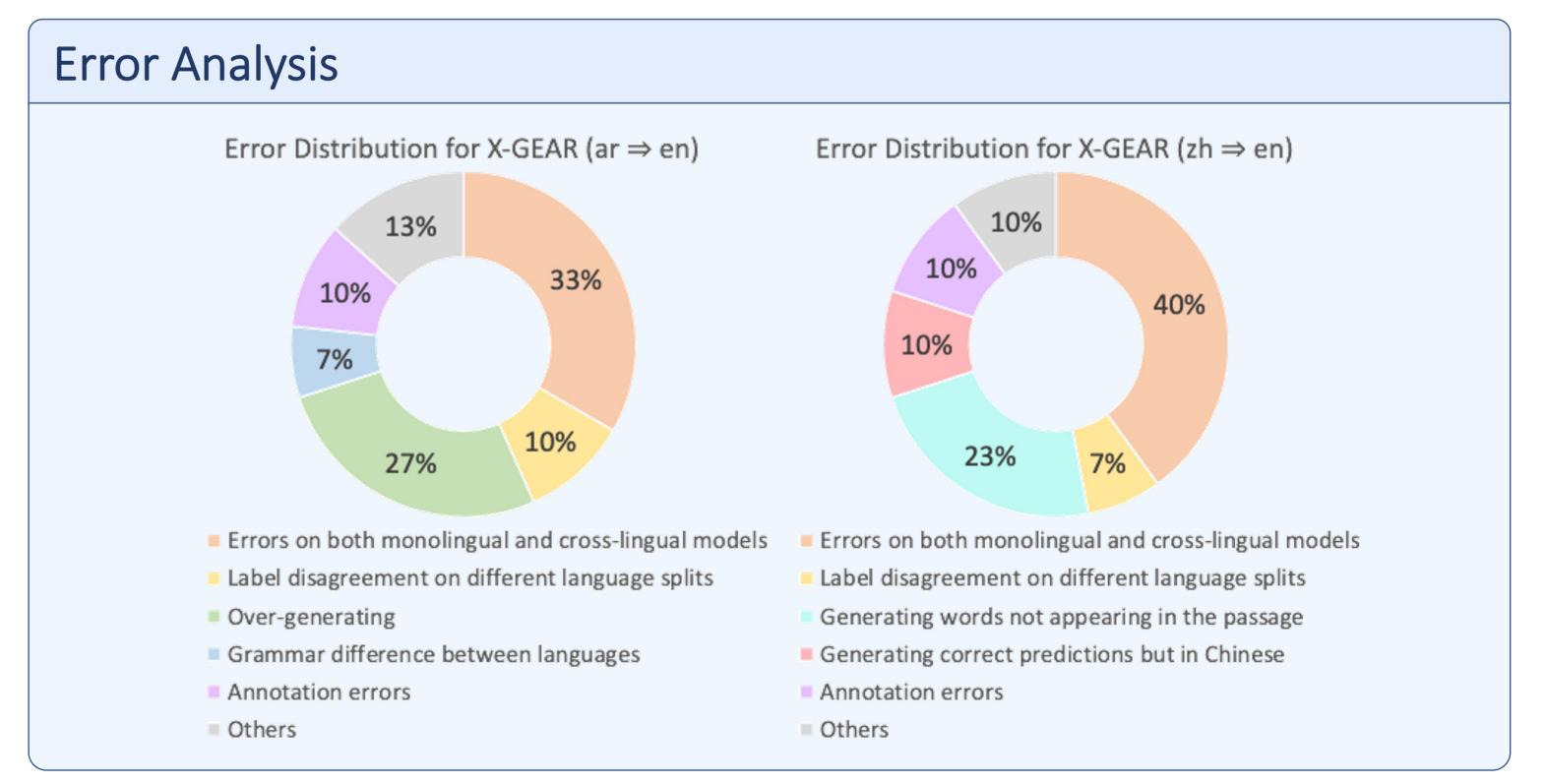
Model	# of parameters	en ↓↓ en	en ↓ zh	en ↓ ar	ar ↓↓ ar	ar ↓ en	ar ↓↓ zh	zh ↓↓ zh	zh ↓ en	zh ↓ ar	avg
OneIE (XLM-R-large) (Lin et al., 2020) CL-GCN (XLM-R-large) (Subburathinam et al., 2019) GATE (XLM-R-large) (Ahmad et al., 2021) GATE (mBART-50-large) GATE (mT5-base)	~570M ~570M ~590M ~630M ~590M	63.6 59.8 67.0 65.5 59.8	42.5 29.4 49.2 43.0 47.7	37.5 25.0 44.5 38.9 32.6	57.8 47.5 59.6 58.5 45.4	27.5 25.4 27.6 27.5 20.7	31.2 19.4 26.3 26.1 21.0	69.6 62.2 <b>70.6</b> 65.9 64.0	51.5 40.8 46.7 45.3 35.3	31.1 23.3 <b>37.3</b> 30.2 22.8	45.8 37.0 47.6 44.5 38.8
TANL (mT5-base) (Paolini et al., 2021)	∼580M	59.1	38.6	29.7	50.1	18.3	16.9	65.2	33.3	18.3	36.6
X-GEAR (mBART-50-large) X-GEAR (mT5-base)	~610M ~580M	68.3 67.9	48.9 <u>53.1</u>	37.8 42.0	59.8 66.2	30.5 27.6	29.2 30.5	63.6 69.4	45.9 <u>52.8</u>	32.3 32.0	46.2 49.1
X-GEAR (mT5-large)	∼1230M	71.2	54.0	44.8	68.9	32.1	33.3	68.9	55.8	<u>33.1</u>	51.3

Alation Study on Copy Mechanism

Model	en ↓ xx	ar ↓ xx	zh ↓ xx	xx ↓ en	xx ↓ ar	xx ↓ zh	avg
mBART-50-lar	ge   <b>51.6</b> 50.9	39.8 <b>42.2</b>	47.2 <b>49.6</b>	48.2 <b>50.6</b>	43.2 <b>43.5</b>	47.2 <b>48.7</b>	46.2 <b>47.6</b>
mT5-base - w/o copy	54.3 52.1	<b>41.4</b> 39.5	<b>51.4</b> 47.6	<b>49.4</b> 48.1	<b>46.7</b> 42.7	<b>51.0</b> 48.5	<b>49.1</b> 46.4
mT5-large - w/o copy	<b>56.7</b> 55.1	44.8 <b>45.0</b>	<b>52.6</b> 51.5	<b>53.0</b> 52.0	<b>48.9</b> 46.3	52.1 <b>53.2</b>	<b>51.3</b> 50.5

Including event type in prompts

Five Iraqi civilians, including a woman, were killed <Type> Event-type <Trigger> killed <Template> <Agent> Monday when their houses were hit by a missile fired <SEP> [None] </Agent> <Victim> [None] </Victim> <Instrument> by the US - led coalition warplanes, witnesses said. [None] XX XX XX Model XX XX XX en X-GEAR (mT5-base) | 54.3 | 41.4 | 51.4 | 49.4 | 46.7 | 51.0 | 49.1 53.3 39.3 **52.3** 49.2 46.5 49.2 48.3 w/ English Tokens w/ Translated Tokens | 51.7 | 40.4 | 52.2 | 49.8 | 45.6 | 48.8 | 48.1 w/ Special Tokens 52.3 39.7 51.8 49.0 45.4 49.3 47.9



## Conclusion

- We propose X-Gear, a generation-based model for zero-shot cross-lingual event argument extraction

  X-Gear
  - Inherit the benefits of generation-based models
  - Language-agnostic templates
  - Copy mechanism
- Significant improvements over previous baselines

### Code is available at

https://github.com/PlusLabNLP/X-Gear https://github.com/PlusLabNLP/TextEE

