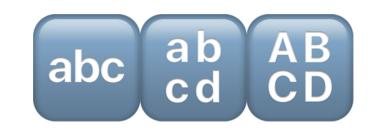




The ELCo Dataset: Bridging Emoji and Lexical Composition









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http://github.com/WING-NUS/ELCo

*All authors thank Prof Min-Yen Kan for his advising.



Victor Li 5:00 PM



Happy Lunar New Year of the Dragon! 🎒 🎉



Dear Friends, Group Mates and WING Alumni @everyone,

- Wishing all the group mates and alumni a fantastic start to the new year! This year, let the Dragon's legendary strength, wisdom, and luck bring us luck, 🝀 success in our studies 😉, achievements in our careers 🚀, and joy in every moment 🤯.
- Let's make this year our best one yet! Go team! 💢 💆 💆



What do they mean?

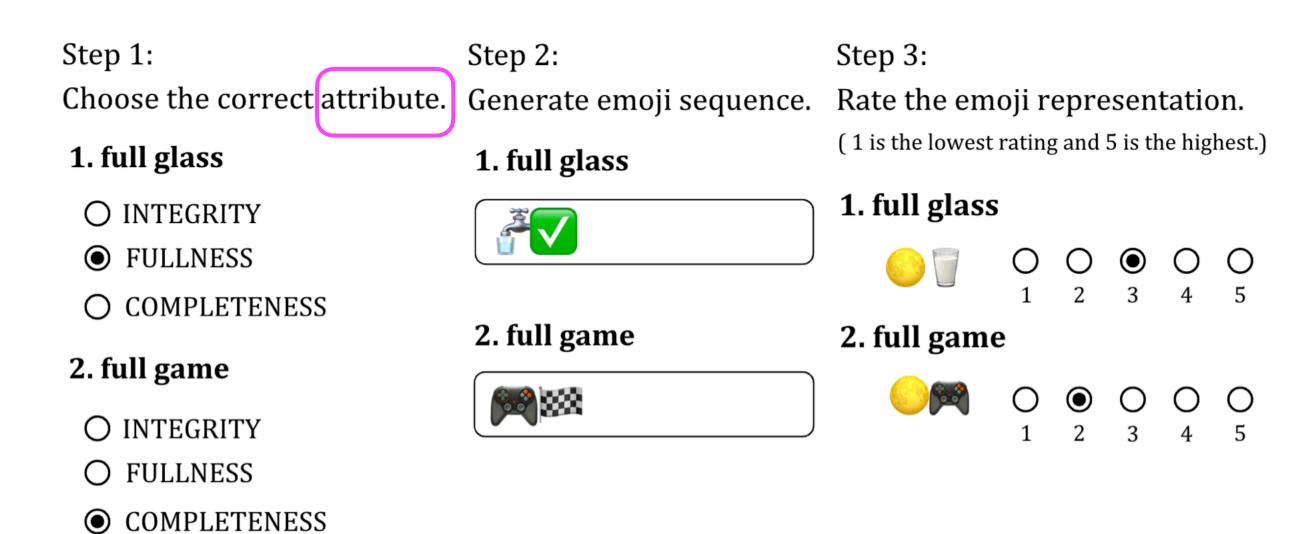
New Problem: Emoji Composition

In this paper, we address:

- Dataset;
- Structure;
- Meaning;

ELCo Dataset Creation

Annotation



Annotation workflow: ELCo's annotation process consists of three steps: (1) select the attribute of the phrase, (2) execute the annotation, and (3) rate the output from a rule-based system, Emojinating.

- HeiPLAS dataset: We choose 209
 AN compounds encompassing
 45 adjectives and 77 attributes.
- 40 university students (IRB approved).
- 1,655 responses received.

ELCo Dataset Creation

Validation

English Phrase	Attribute	ELCo's Annotations (length = 2)	ELCo's Annotations (length > 2)	Average length of ELCo's Annotation	Emojinating's Output	Emojinating Rating
full attention	INTEGRITY			2.43		3.1
full glass	FULLNESS	🥛 🚟 and 💣✓		2.71		3.0
full game	COMPLETENESS	⋒ ⋙ and ⋒ <u></u>	100 FREE	2.29		2.7
full auditorium	FULLNESS	満論	🗎 🚇 🧰 and 🔤 😇 😇 🤀 🛅	4.14	○ 🛵	2.0
full life	FULLNESS	○ ♥	♥‱ š š š 2 and ≥6	4.00		1.1

Compared with Emojinating:

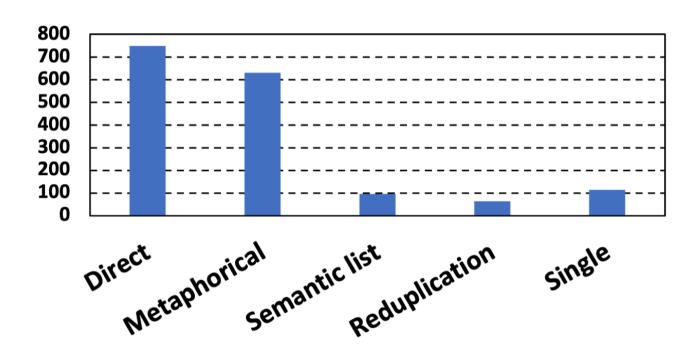
- ELCo's lengths?
- Is ELCo being literal?
- Is ELCo more metaphorical?

Corpus study: Structures for Emoji Composition

6

	Compositional Strategy	EN Phrase	Emoji Sequence
Ex. 1	Direct	right man	√ 😇
Ex. 2	Metaphorical	right man	
Ex. 3	Metaphorical	clear explanation	<u>100</u> *
Ex. 4	Metaphorical	fresh bread	
Ex. 5	Semantic list	bright future	
Ex. 6	Reduplication	big group	ňň ňň ňň
Ex. 7	Single	right thing	✓

Example for compositional structures.

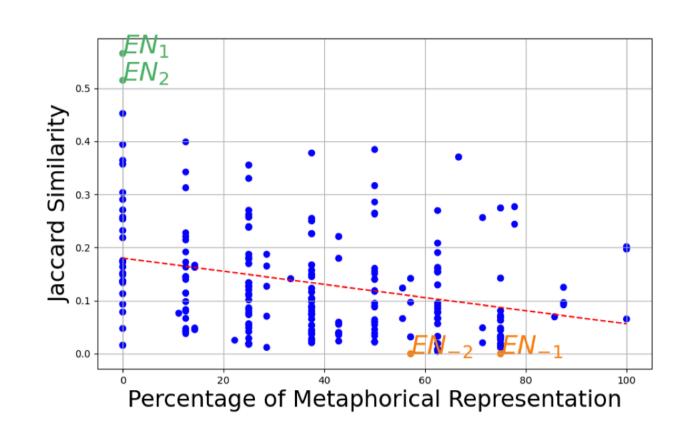


Corpus Study: Number of compositional structures identified in our corpus study (1,655 samples in total).

Inspired by Cohn's emoji grammar, following structures are identified:

- Direct: Translate;
- Metaphorical: Embody;
- Semantic list: Imply;
- Reduplication: Intensify;
- Single: Essence.

Corpus study: Structures for Emoji Composition

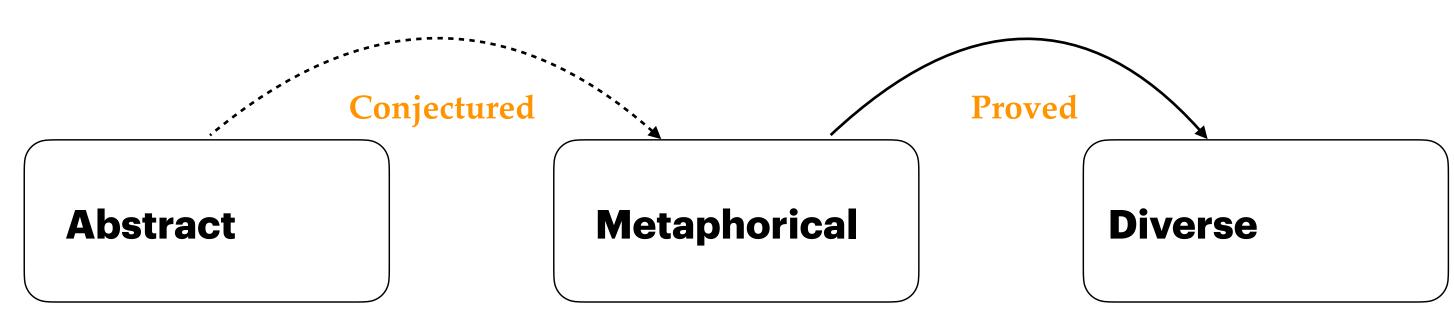


Jaccard Similarity: $J(A,B) = \frac{|A \cap B|}{|A \cup B|}$

• k = -0.123.

\overline{ID} : EN PHRASE	Human Annotation Samples	Main Compositional Strategy	Jaccard Similarity
EN_1 : WRONG MEDICINE	X0, 😭 0, X00, 🗟 0	Direct	0.57
EN_2 : WRONG ROAD	X , & , , & , , & , ,	Direct	0.51
EN_{-2} : FAR SIDE	∦ 🚱 ∦ , 🔼 🚙 , 🕞 👉 , 🛇 式 📏	Metaphorical	0.0
EN_{-1} : IMMEDIATE INFLUENCE	//// ★ Y // // // // // // // // // // // // /	Metaphorical	0.0

More diverse.



Formalization

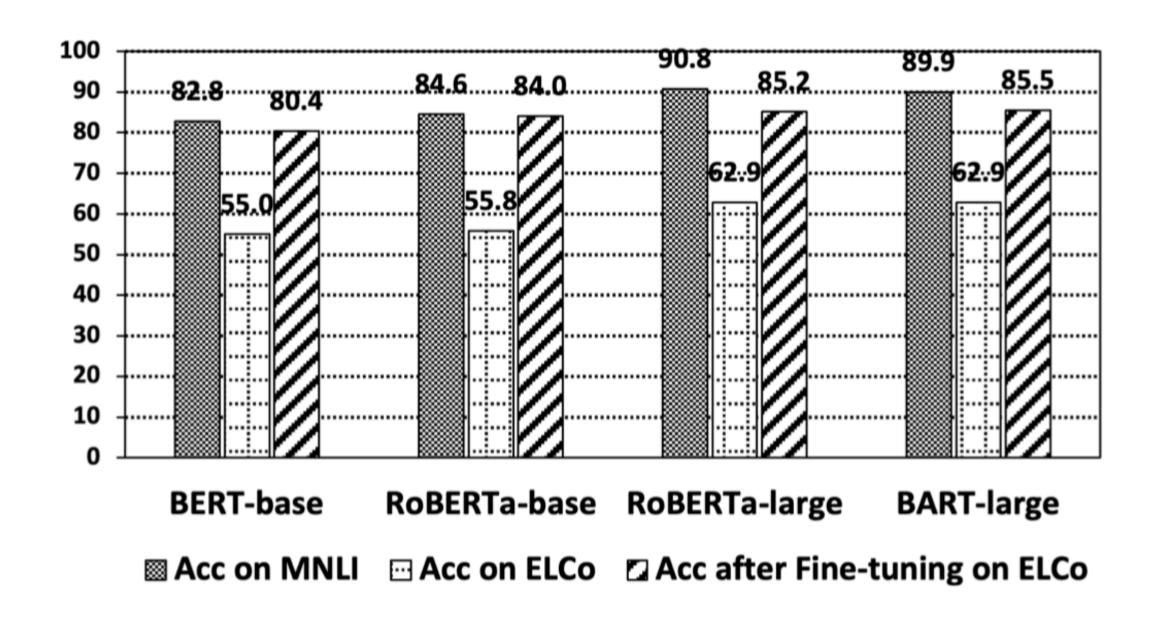
Determines if a sequence of emojis *EM* (em₁ em₂ ... em_n) implies a English phrase *EN* (en₁ en₂ ... en_n). Formally:

- Premise: P em₁ em₂ ... em_n

Input	Golden Label	
Premise: This is 🧐🚣.	Entailment	
Hypothesis: This is full attention.	Linamien	
Premise: This is 📆.	Non-Entailment	
Hypothesis: This is full attention.	Mon-⊑ntaiiment	

- Dataset split: Roughly 70:15:15 for training, validation, and testing.
- Sampling:
 - Noun flipping by Shwartz and Dagan (2019): AN -> AN' -> EM'
 - (AN, EM) and (AN, EM')
- Models: BERT, Roberta, BART, ChatGPT-3.5.

Overall performance



- Finding 1: Comparing with traditional NLI?
- Finding 2: Does fine-tuning help?

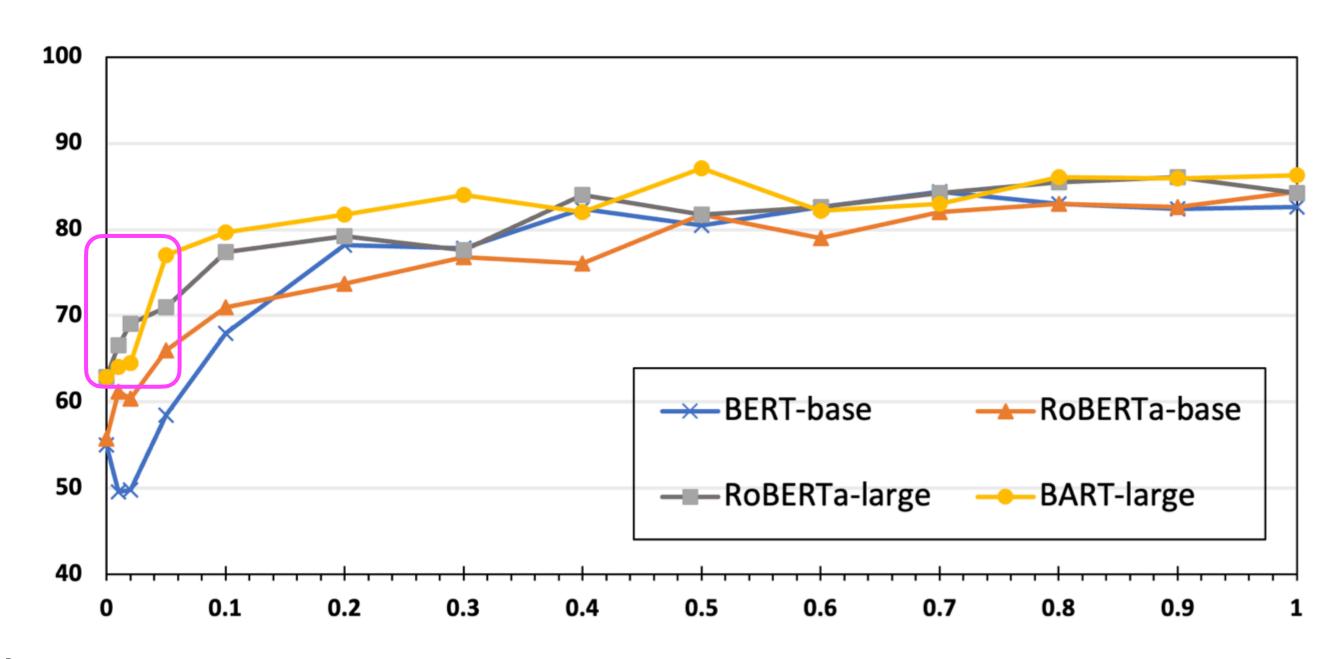
Fine-grained analysis

w/o fine-tuning on ELCo

Fine-tuned on ELCo

	BERT base ROBERTabase ROBERTalarge AV9				BERTbase ROBERTabase ROBERTalarge BARTlarge AVG					
	BERTbase	ROBELL	ROBELL	BART large	Ma	BERTbase	ROBEIL	ROBELL	BART large	Ma
Direct	34.7	35.6	41.5	51.7	40.9	85.1 _{2.3}	89 _{3.3}	91.9 _{2.2}	88.5 _{2.8}	88.6
Metaphorical	19.4	24.7	34.4	36.6	28.8	68.4 _{3.9}	73.1 _{2.6}	80.4 _{1.9}	82.8 _{4.3}	76.2
Semantic list	33.3	41.7	50.0	58.3	45.8	86.7 _{7.5}	$78.3_{4.6}$	$85.0_{3.8}$	91.7_{0}	85.4
Reduplication	13.3	0	6.7	0	<u>5.0</u>	65.4 _{3.0}	52.0 _{7.3}	$62.7_{6.0}$	$88.0_{8.7}$	<u>67.0</u>
Single	19.0	19.0	19.0	52.4	<u>27.4</u>	66.7 _{3.4}	88.6 _{2.6}	85.7 _{4.8}	83.8 _{2.6}	<u>81.2</u>
Negative	83.4	83.0	90.3	82.2	84.7	84.3 _{4.1}	87.2 _{1.9}	85.2 _{0.7}	84.8 _{1.2}	85.4
Overall	55.0	55.8	62.9	62.9	59.2	80.4 _{1.5}	84.0 _{0.8}	85.2 _{0.9}	85.5 _{0.9}	83.8

What's challenging?

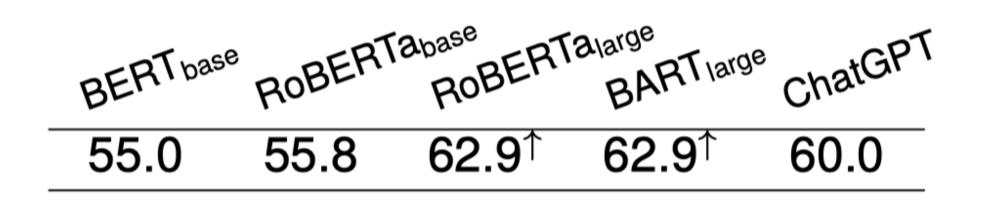


- Can models converge?
- What models converge faster?

Emoji-based Textual Entailment (EmoTE) Task Case study

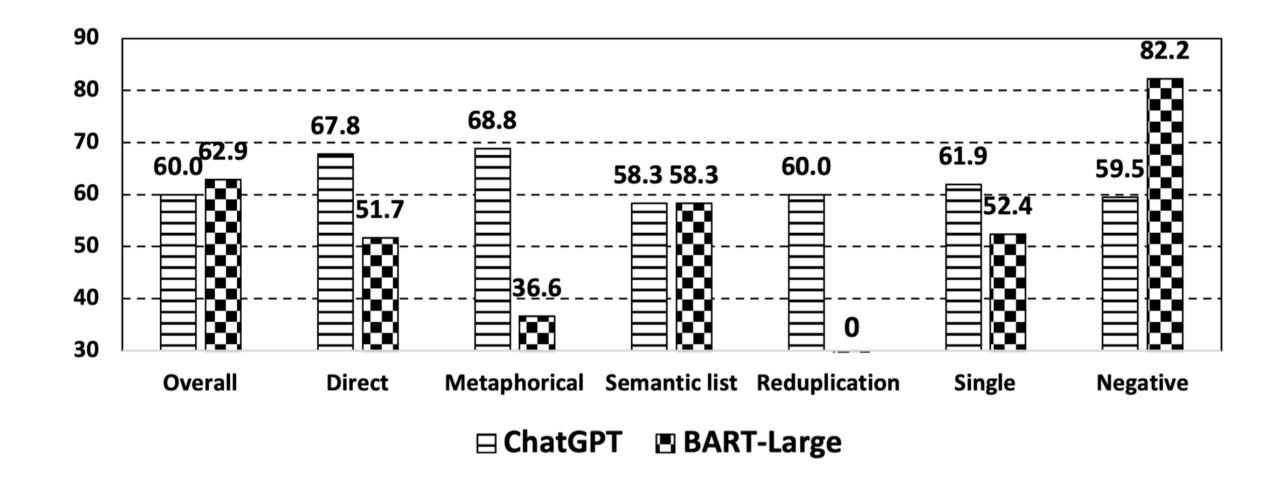
English	Emoji	Pre	Post	
1 big group		×	√	
2 big city		×	\checkmark	
3 hot forehead		×	\checkmark	
4 thin soup	08 08 4	×	\checkmark	
5 big city		×	×	Visual information of emoji.
6 ineffectual ruler		×	×	Commonsense knowledge.
7 full attention		×	×	Distant from EN phrase's meaning.
8 full life		×	× —	Distant from EN phrase's meaning.

ChatGPT



Overall Performance comparison of ChatGPT

- Is ChatGPT competitive?
- What is ChatGPT good at?

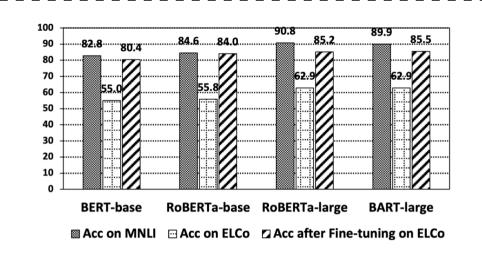


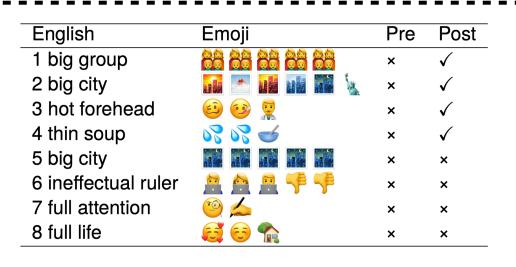
Fine-grained Performance comparison of ChatGPT

Conclusion

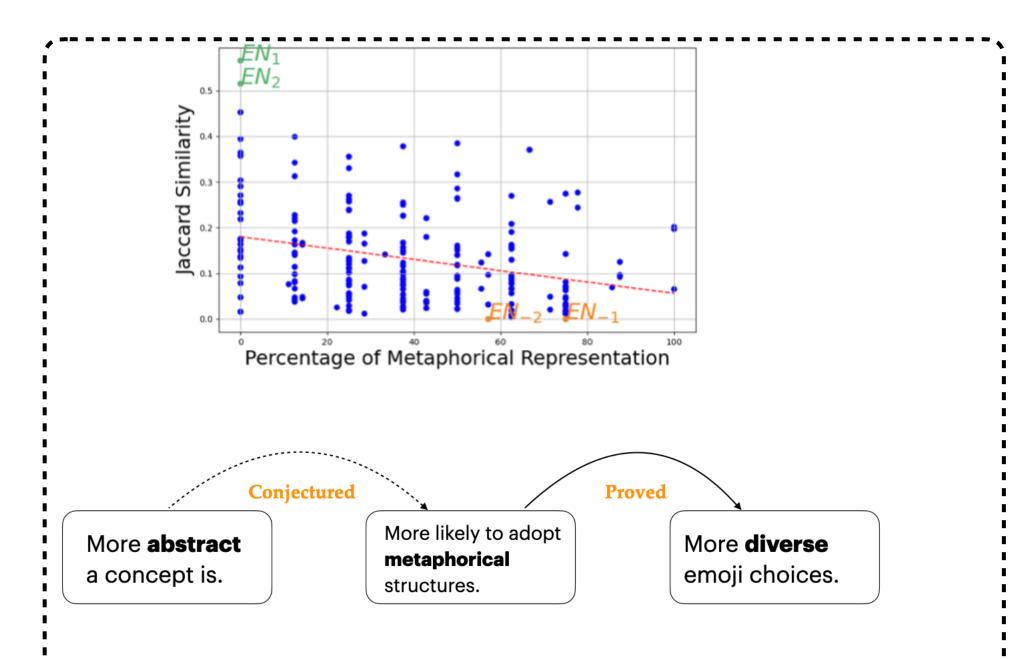
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full life	FULLNESS	○♥	♥ ‱ š š <u>\$</u> 2 and ≥	4.00	O	1.1

ELCo dataset is comprised of 1,655 annotations of 209 EN phrases 45 adjectives and 77 attributes.





EmoTE is challenging for all models, but fine-tuning on ELCo helps to learn useful emoji composition skills.



Corpus study reveals five structures to compose emoji compositions, and we show metaphorical structures use more diverse emojis.

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