

# Compositionality and Generalization in Emergent Languages

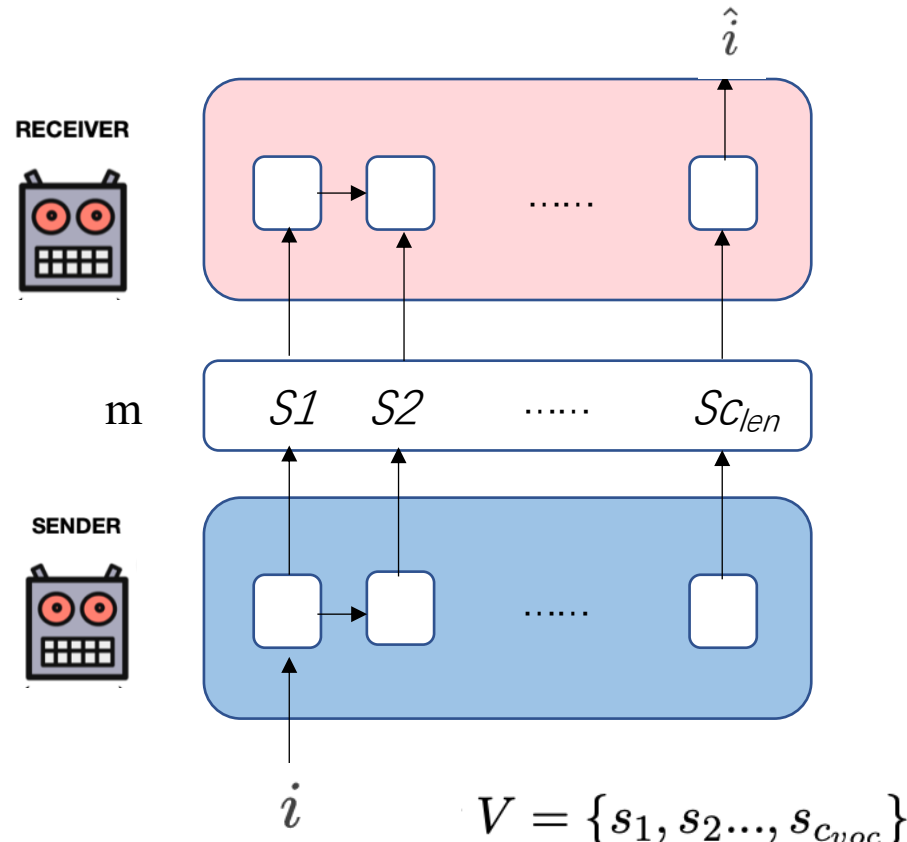
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# Introduction

- Compositionality in human languages
  - Efficient strategy where we need to only encode finite set of words & systematic rules to encode infinite meanings.
  - Enables generalization to novel meanings.
- Compositionality in emergent languages
  - To develop communicating neural network agents, we need to enable them to communicate with compositional languages.
  - Neural network agents are general-purpose learners that we can intervene on to assess what affects the degree of compositionality of their languages. This makes the study of emergent communication important for **language evolution**.

# Replication process -- understanding the game

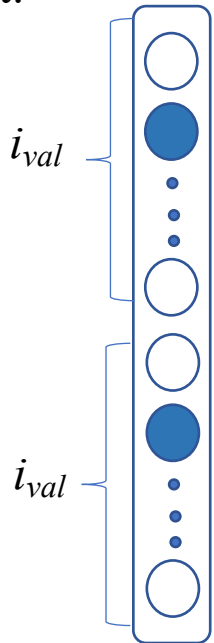
- Sender network receives one input  $i$  and chooses a sequence of symbols from its vocabulary  $V = \{s_1, s_2, \dots, s_{c_{voc}}\}$  of size  $c_{voc}$  to construct a message  $m$  of fixed length  $c_{len}$ .
- Receiver network consumes  $m$  and outputs  $\hat{i}$ .
- Agents are successful if  $\hat{i} = i$  (i.e., Receiver reconstructs Sender's input).



# Replication process -- understanding the game

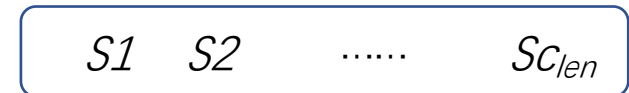
- Each input consists of  $i_{att}$  attributes, each with  $i_{val}$  possible values.
- Messages are of a fixed length  $c_{len}$ . At each position  $j$ , a symbol  $s_j$  is sampled from Sender's vocabulary of size  $c_{voc}$ .

Input:



Example of an input  $i$  with  $i_{att}=2$ . It encodes the second value of the first attribute and the first value of the second attribute.

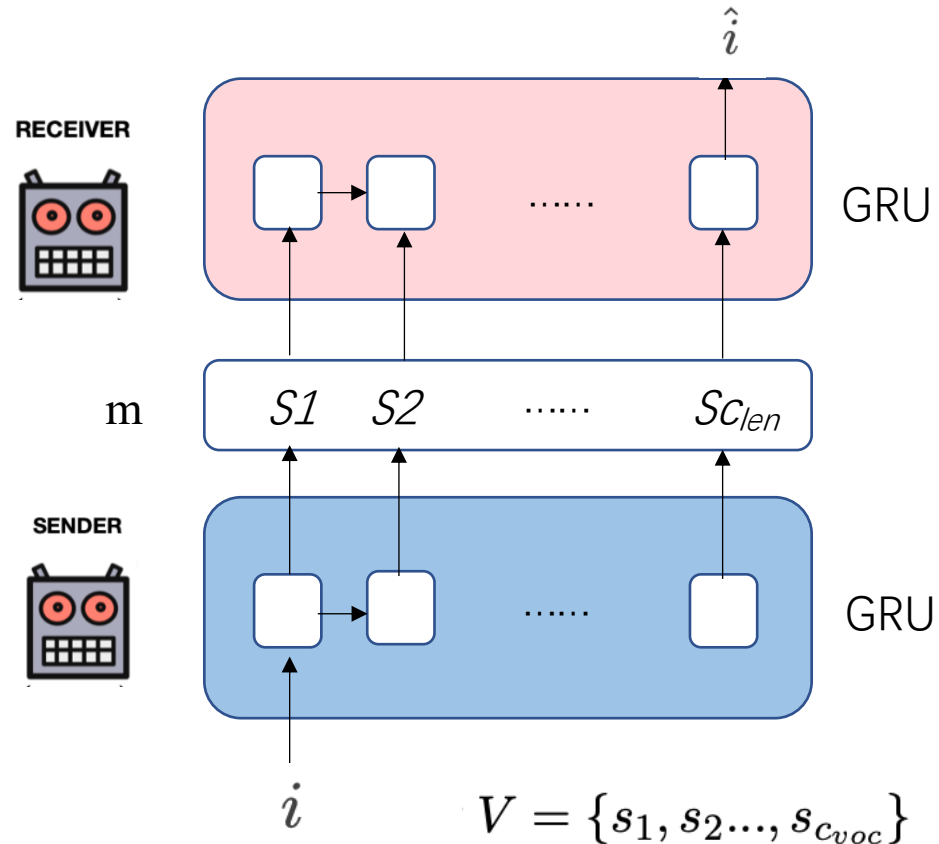
Message:



$$V = \{s_1, s_2, \dots, s_{c_{voc}}\}$$

# Replication process -- understanding agent architecture

- Both agents are implemented as single-layer GRU cells.
  - hidden states of size 500
- Loss: the average cross-entropy between output distributions and Sender's input.



# Replication process

-- Generalization in emergent languages

- Generalization emerges “naturally” if the input space is large.
  - Input size  $|I| = i_{val} i_{att}$
  - Channel capacity  $|C| = c_{voc} c_{len}$
- Sender:  $|C| \geq |I|$
- Parameters:
  - `--n_values=xx --n_attributes=xx --vocab_size=10 --max_len=6 --batch_size=5120 --data_scaler=60 --n_epochs=3000 --sender_hidden=500 --receiver_hidden=500 --sender_entropy_coeff=0.5 --sender_cell="gru" --receiver_cell="gru" --lr=0.001 --receiver_emb=30 --sender_emb=5 --random_seed=9`

$(i_{val}, i_{att}) \backslash \begin{matrix} c_{voc} \\ c_{len} \end{matrix}$	5				10				50				100			
	2	3	4	{6,8}	2	3	4	{6,8}	2	3	4	{6,8}	2	3	4	{6,8}
(4,4)			X	X		X	X	X	X			X	X			X
(5,2)	X	X	X	X	X	X	X	X	X			X	X			X
(5,3)		X	X	X		X	X	X	X			X	X			X
(5,4)			X	X		X	X	X	X			X	X			X
(10,2)		-	X	X	X	X	X	X	X			X	X			X
(10,3)				X		-	X	X	X	X	X	X	X			X
(10,4)				{-, X}			-	X		X	X	X	-	X	X	X
(16,2)			-	X		X	X	X	X			X	X			X
(25,2)			-	X		-	X	X	X			X	X			X
(50,2)				X			-	X	-	X	X	X	X	X	X	X
(100,2)				{-, X}			-	X		X	X	X	-	X	X	X

Table 3: Grid search. ‘X’ indicates tested settings with at least one successful run. ‘-’ indicates tested settings without any successful run. Finally, blank cells correspond to settings that were not explored for the reasons indicated in the text.

# Replication process -- Generalization in emergent languages

See replication codes & files: [https://github.com/yuqing0304/compositionality\\_replicate](https://github.com/yuqing0304/compositionality_replicate)

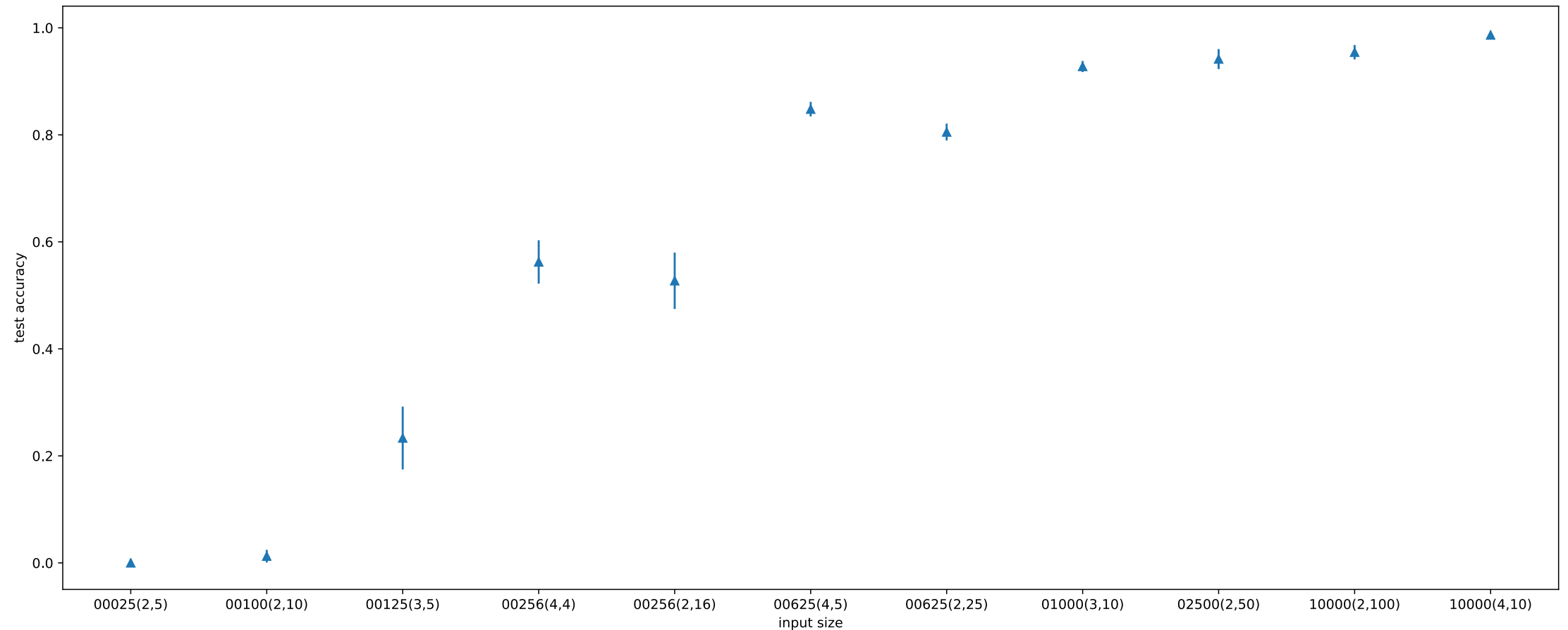
Test accuracy:

```
{ '00025(2,5)': { 'acc_mean': 0.0, 'acc_sem': 0.0 },  
'00100(2,10)': { 'acc_mean': 0.0125, 'acc_sem': 0.011858541225631423 },  
'00125(3,5)': { 'acc_mean': 0.233333334028720856, 'acc_sem': 0.05868939128794347 },  
'00256(2,16)': { 'acc_mean': 0.527272741496563, 'acc_sem': 0.05269591837361792 },  
'00256(4,4)': { 'acc_mean': 0.5625, 'acc_sem': 0.04050462936504913 },  
'00625(2,25)': { 'acc_mean': 0.8052631676197052, 'acc_sem': 0.015779727972030193 },  
'00625(4,5)': { 'acc_mean': 0.8479999780654908, 'acc_sem': 0.013623509400751192 },  
'01000(3,10)': { 'acc_mean': 0.92777778506279, 'acc_sem': 0.010356306263776276 },  
'02500(2,50)': { 'acc_mean': 0.941666716337204, 'acc_sem': 0.018680427008010262 },  
'10000(2,100)': { 'acc_mean': 0.9543877124786377, 'acc_sem': 0.013494911879231615 },  
'10000(4,10)': { 'acc_mean': 0.9865853190422058, 'acc_sem': 0.004432800984419676 } }
```

acc\_mean: mean accuracy (10 different initializations per setting).

acc\_sem: standard error of the mean.

# Replication process -- Generalization in emergent languages



Emergent languages are able to almost perfectly generalize to unseen combinations as long as input size  $|I|$  is sufficiently large.



# My reflection:

- The code for the EGG framework is heavily encapsulated, so it takes time for me to understand its functioning.
- The hyperparameters used in the paper are needed in order to replicate the same results. It may take a long time to try different hyperparameters (I have tried replicating the experiment mentioned in *Anti-efficient encoding in emergent communication* first. But I haven't replicated the results yet after trying many parameters.)
- I think these experiments are innovative and insightful and I love to play these games, though it takes time for me to be skilled.