

# **Emergent Quantized Communication**



The Technion – Israel Institute of Technology

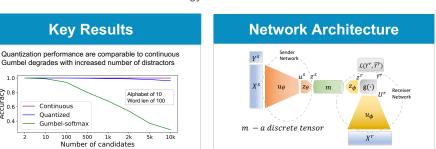
1.0

0.8 ي

0.4

10

Accur 0.6



## **The Problem**

- Communication emerges from letting artificial agents solving tasks Assess its characteristics
- Constraining the channel
   Discrete communication
- NN need to be continuous for gradients to flow Reinforcement learning
  Gumbel softmax
- Quantization as a superior discretization approach
- Can be optimized end-to-end
  Resulted in good performance



The Channel				Baselines
	Continuous           Floating point           0.10         0.23         0.69           0.10         0.23         0.69           0.72         0.54         0.21           hannel: Single-wort channel: Multipl         0.01/10			Continuous Communication: • Good performance with sufficient capacity • Enables end-to-end training with back- propagation Gumbel-softmax: • A continuous approximation for a categorical distribution • Discrete messages are approximated via sampling procedure
The Algorithm				Main Idea and Running Example
Algorithm 1: Quantizing continuous communication. 1: msg: continuous message > Floating point vector				Scales continuous messages to alphabet size

#### mg: continuous message ▷ Floating point vector ▷ Setthe alphabet range procedure NORMALTE(msg) min.elem - min(msg.elements) maz.elem - max(msg.elements) masg (- msg.elems - min.elem)/maz.elem return msg end procedure and rounds values to the closest integer Example for alphabet of 4 and word length of 5 Initial Msg -3.5 0.2 1.3 -1.3 2.7 $\begin{array}{l} d \operatorname{procedure} \\ occdure (UANTLE(msg)) \\ msg \leftarrow Normalize(msg) \\ s, msg \leftarrow msg(sec, msg) \\ s, msg \leftarrow msg(sec, msg) \\ s, msg \leftarrow msg(sec, msg) \\ degt, msg \leftarrow degt, msg \\ degt, msg \in degt, msg \\ degt, msg \in degt, msg \\ degt, msg \in degt, msg \\ s, scale back \\ return (dedt, msg, discrete, msg) \\ sg \leftarrow Quantize(msg) \end{array}$ 0 3.7 4.8 2.2 6.2 Normalize 0 0.59 0.77 0.35 1.0 Scaling by 4 0 2.38 3.09 1.4 4.0 Rounding 0 2.0 3.0 1.0 4.0 Scaling back 0 0.5 0.75 0.25 1.0 Rcvr

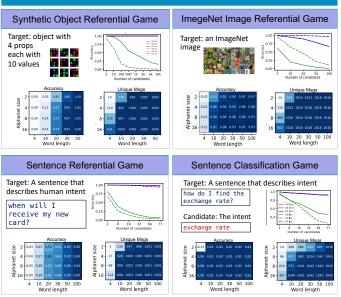
#### Explanation and Pytorch Implementation

- quantize\_per\_tensor and dequantized methods allow gradients to flow through non differentiable rounding operation
- Implementing the straight through estimator during backward by overriding the backward methods and returning the input gradients

### The Role of Channel Capacity

- Plays significan role in agens ability to accomplish a task but is not the only factor
- Sometimes even unlimited channel capacity is not enough
  - · Continuous channel has unlimited capacity, still cannot perfectly solve the classification game • With RNN even a Gumbel-softmax channel has enough capacity to solve the e.g., object game
- Optimization plays significant role too
- · Limiting the capacity of a quantized channel can be achived by controllling word leangth and/or alphabet size

## Games Datasets and Results



#### Classification and Referential Games

- In the referential game the message sent by the sender needs to accurately describe the target.
- In classification game on the other hand, message need to let parties agree on the object's class.

## **Open Issues**

- · Which inductive biases needs to be added beyond merely requiring the communication to be discrete?
- · Can a non-natural, yet discrete, language that emerge from agents' communication be translated to natural language?
- Do humans represent the world using continuous or discrete elements, and when during the communication process discretization occur?

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