

# Challenges and research directions in Neural Machine Translation

### multilingual, unsupervision, fairness

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### Words Embeddings



### **Contextual Words Embeddings**



### **Contextual Words Embeddings**



### **Contextual Word Embeddings use Transformers**



### **Neural Machine Translation**



Los pájaros pueden volar

### Three different architectures in 2 years

#### 

#### **RNN WITH ATTENTION (2015)**

### Three different architectures in 2 years

# RNN WITH ATTENTION (2015)



### Three different architectures in 2 years

#### **RNN WITH ATTENTION (2015)**





#### **TRANSFORMER (2017)**



## **Big Successes**

### **Google translation Evolution**



### Microsoft claims human parity

### Achieving Human Parity on Automatic Chinese to English News Translation

Hany Hassan, Anthony Aue, Chang Chen, Vishal Chowdhary, Jonathan Clark,
Christian Federmann, Xuedong Huang, Marcin Junczys-Dowmunt, William Lewis,
Mu Li, Shujie Liu, Tie-Yan Liu, Renqian Luo, Arul Menezes, Tao Qin,
Frank Seide, Xu Tan, Fei Tian, Lijun Wu, Shuangzhi Wu, Yingce Xia,
Dongdong Zhang, Zhirui Zhang, and Ming Zhou

Microsoft AI & Research

### Is Machine Translation Solved?

### Quality in Machine Translation depends on training data



### Quality in Machine Translation depends on training data



# **Research Directions**

Multilingual systems & Unsupervised systems

### Low-resourced languages can benefit from high-resourced

**Universal Encoder-Decoder** 



#### Low-resourced languages can benefit from high-resourced

**Universal Encoder-Decoder** 

Language-Specific Encoder-Decoders



### Low-resourced languages can benefit from high-resourced

#### **Universal Encoder-Decoder**

Shared Vocabulary

- ✓ Zero-shot
- Transfer learning from high-resourced languages to low-resourced (with the same script)
- x Detrimental for high resourced languages

- Language-Specific Encoder-Decoders
  - Independent vocabulary

- ✓ Zero-shot
- Incremental training of new languages and domains

x No transfer learning from high-resourced to low-resourced (with the same script)

#### Machine Translation can be trained on unlabelled data $L_1$ $L_2$ **Monolingual Data Monolingual Data** $D_2$ $D_1$ hello mauvaise bon mon -good - my\* bonjour Cross-lingual embedding Word embedding Word embedding $E_{1}/E_{2}$ $L_{1}/L_{2}$

# Questioning our data...

### **Biased Translations**



# Bad Translations can generate Automated Bias



# **Towards Fairer Systems**

Data augmentation & Debiasing algorithms

### Data augmentation

#### **Biased Text**

<Malik:he> is an aspiring singer who works as a salesman in a car showroom. One day he meets 
<Sonia:she> Saxena daughter of Mr. Saxena when goes to deliver a car to home as birthday present

#### **Debiased Text**

**Malik:she>** is an aspiring singer who works as a salesman in a car showroom. One day he meets **Sonia:he>** Saxena daughter of Mr. Saxena when goes to deliver a car to home as birthday present





### **Debiased algorithms**



### **Debiased algorithms**



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### Inspiration from other areas... healthcare?

Under-representation of racial and ethnic minorities in clinical trials





### More about our research

# www.talp.upc.edu

# www.costa-jussa.com