



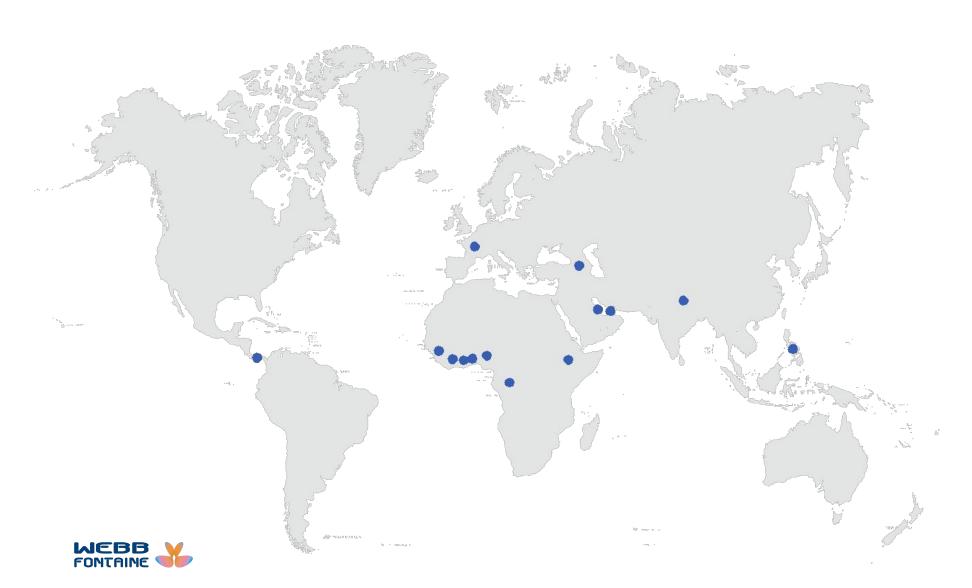
#### **About Us**

We believe in looking beyond barriers to shape tomorrow's international trade. We develop high-tech software and services to benefit all trade stakeholders and make their lives easier.





## **Our Locations**



Armenia

Bahrain

Benin

Congo

**Ivory Coast** 

Ethiopia

France

Guinea

Ghana

Nepal

Nigeria

Panama

Philippines

**United Arab Emirates** 

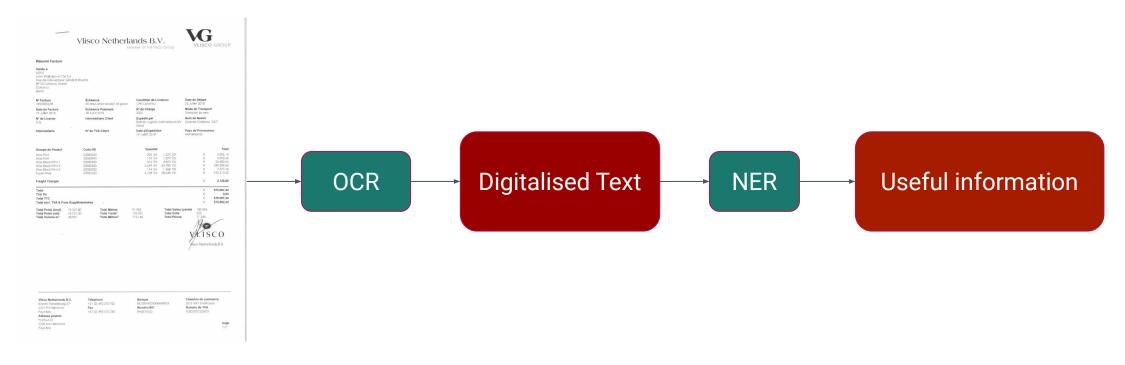
# **Optical Character Recognition (OCR)**





### **Information Extraction From Invoices**

### PDF input





## Named Entity Recognition (NER)

```
the afternoon TIME
 May 10, 2017 DATE, deputy press
secretary Sarah Sanders PERSON
                                 spoke
to the President about his decision to fire
               and then spoke to
 Comey PERSON
reporters in a televised press conference.
```



# **Optical Character Recognition in the Wild**





#### **Street View House Numbers Extraction**





## **Licence Plates Detection and Recognition**





# Optical Character Recognition from PDF documents

While there was wet weather and a risk of flooding in the east and north, parts of the west basked in sunshine.

The weather will gradually improve in the east today before the whole country gets



```
"textblock": [
    "text": "While there was wet Weather\n
        and a risk of flooding in the\n
        east and north, parts of the\n
        west basked in sunshine.\n
        The weather will gradually\n
        improve in the east today\n
        before the whole country gets",
    "left": 0,
    "top": 0,
    "width": 1860,
    "height": 1200
}
```



## **OCR Engines 2019**

Why should one reinvent bicycles?



## **OCR Engines 2019**

# Why should one reinvent bicycles?

- Accuracy
- Price



## **OCR Engines 2019**

# Why should one reinvent bicycles?

- Accuracy
- Price
- Privacy



# **OCR Engine**

#### **Detection**



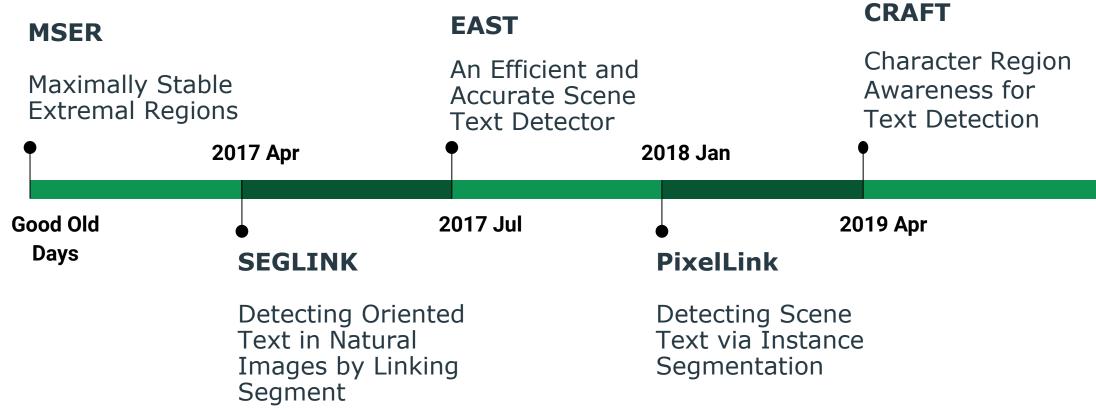
#### Recognition

WAITING

PLEASE
TURN OFF
YOVR
ENGINE

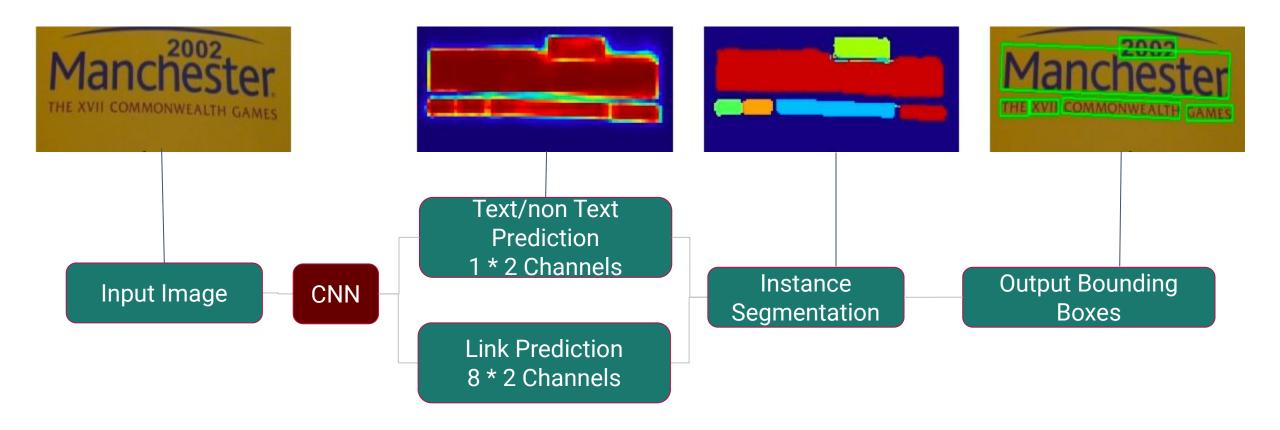


## **Popular Detection Models**



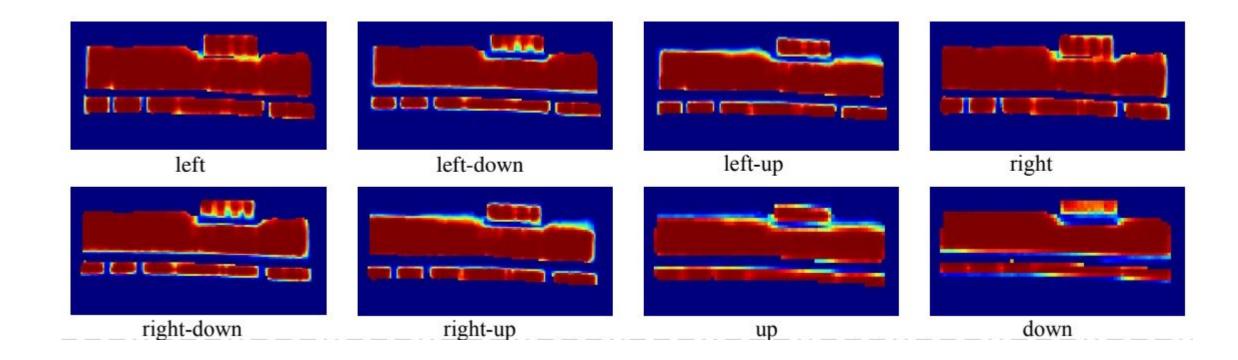


#### Prediction Pipeline



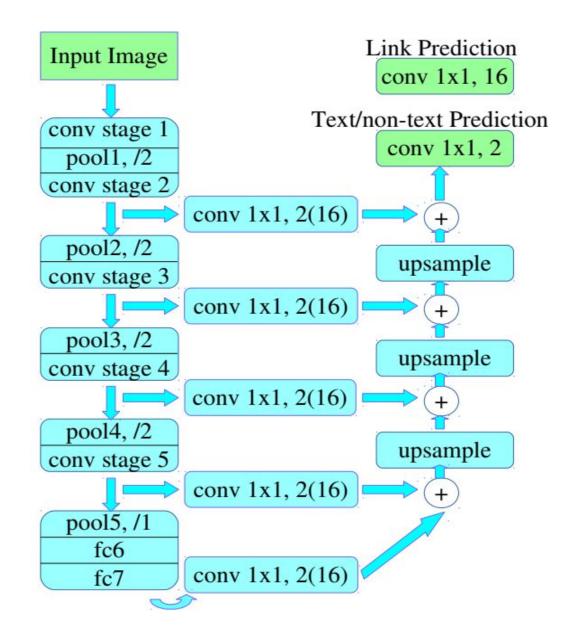


### Link Heatmaps





Model Architecture





#### Loss Function

$$L = \lambda L_{pixel} + L_{link}$$

- L Loss
- ullet L $_{pixel}$  Pixel Loss
- L<sub>link</sub> Link Loss
- λ Importance Coefficient

$$L_{pixel} = \frac{1}{(1+r)S} W L_{pixel\_CE}$$

- *r* negative positive ration
- S number of positive pixels
- W per pixel weight matrix

$$w_i = \begin{cases} \frac{B_i}{S_i}, & if \ pixel \ is \ positive \\ 1, & if \ negative \ pixel \ is \ selected \\ 0, & otherwise \end{cases}$$



#### Loss Function

$$L = \lambda L_{pixel} + L_{link}$$

- L Loss
- L<sub>pixel</sub> Pixel Loss
- L<sub>link</sub> Link Loss
- λ Importance Coefficient

$$L_{link} = \frac{L_{link\_pos}}{rsum(W_{pos\_link})} + \frac{L_{link\_neg}}{rsum(W_{neg\_link})}$$

- ullet  $L_{link\_pos}$  Loss for positive pixels
- L<sub>link\_neg</sub> Loss for negative pixels

$$L_{link\_pos} = W_{pos\_link}L_{link\_CE}$$
$$L_{link\_neg} = W_{neg\_link}L_{link\_CE}$$



## **CRAFT**

#### Prediction map



E|A.I|E.P| 優質電池製造公司總部位於上海 | SHANGHA||ROAD||76242|BUILDING|43

INVOICE

SOCIETE ANISAL

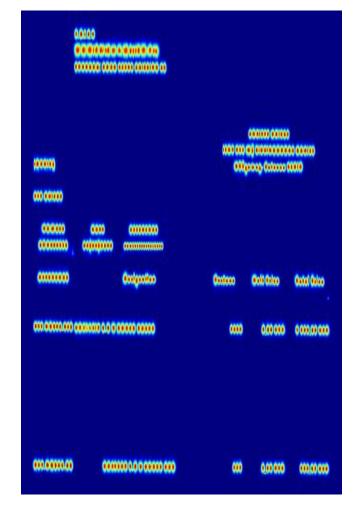
ILOT 643 M/ZINZINDOHOUE DANIEL

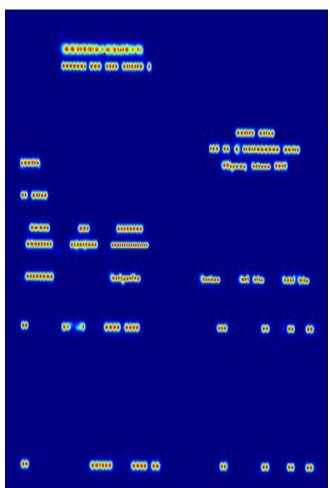
Gbégamey-Cotonou BENIN

STE ANISAL

NUMBER	DATE	REFERENCE				
CHN6578FR	11/07/2017	CHN6578FR11072017				

REFERENCE	Designation	Cartons	Unit Price	Total Price		
AST.NO956.R26 B	ATTERIE 1.2 V COVER PAPER	1650	2,00 USD	3300,00 USD		
AST.NO521.R6	BATTERY 1.2 V COVER PVC	200	2,50 USD	500,00 USD		

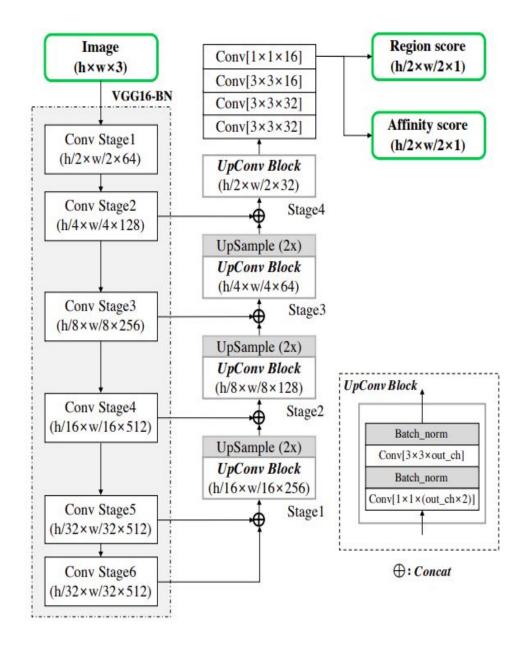






#### **CRAFT**

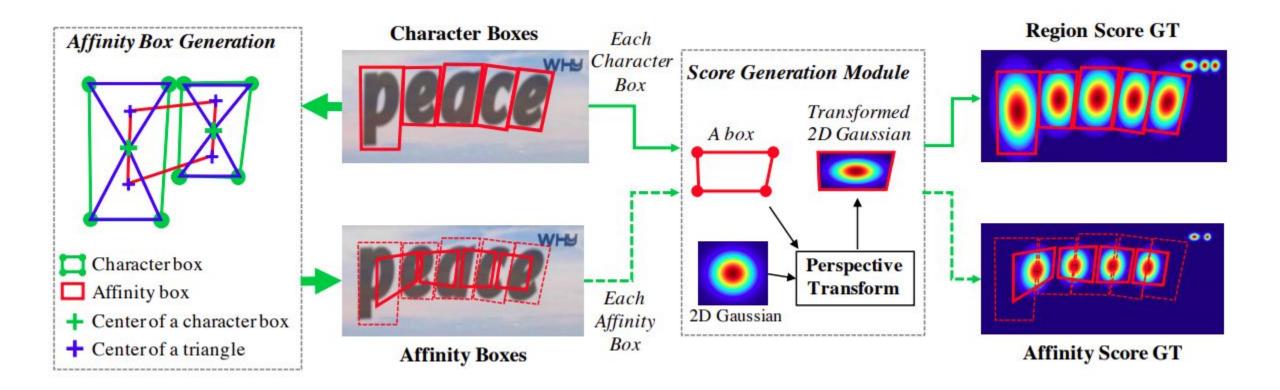
#### Model Architecture





#### **CRAFT**

#### **Data Generation**

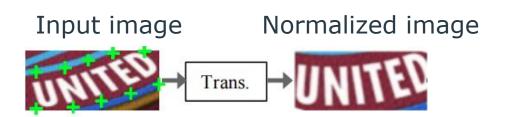




Main Stages

**Transformation** 

Normalization of input image using spatial transformation network (TPS)





Main Stages

**Transformation** 

Feature extraction

Mapping the input image to representation that focuses on attributes that are more relevant to character recognition (VGG, Resnet, RCNN)





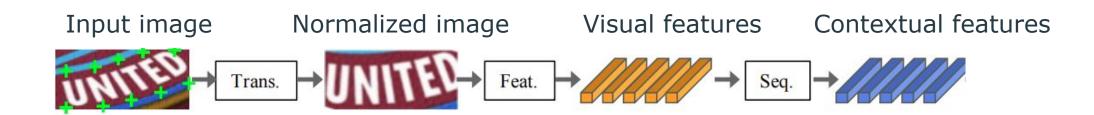
Main Stages

**Transformation** 

Feature extraction

Sequence modeling

Capturing the contextual information within a sequence of characters for the next stage to predict each character more robustly, rather than doing it independently (BiLSTM)





Main Stages

**Transformation** 

Feature extraction

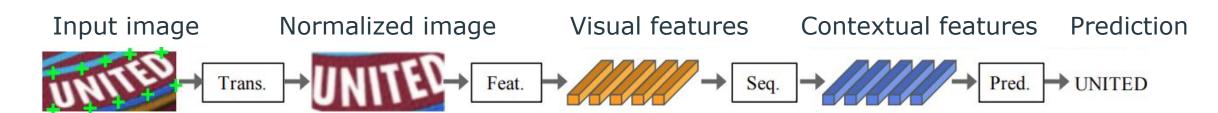
Sequence modeling

Estimating the output character sequence from the identified features of an image

(Attention CTC)

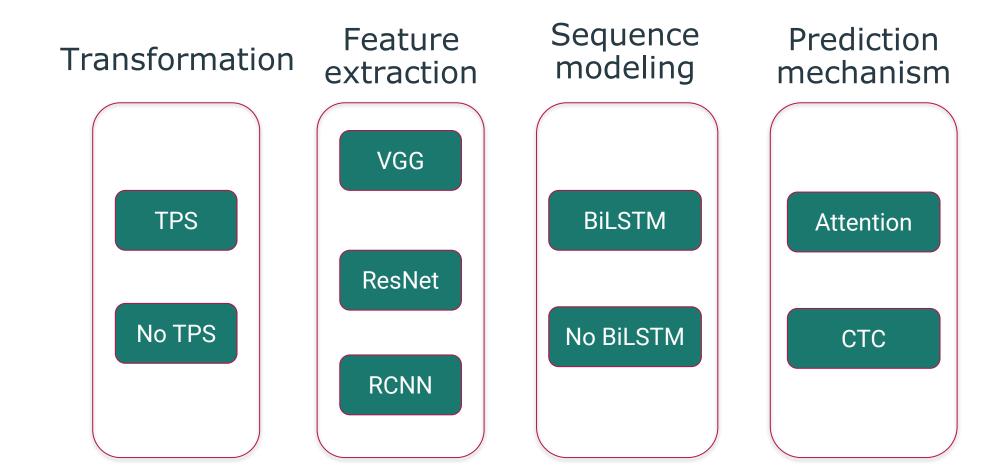
(Attention, CTC)

#### Prediction



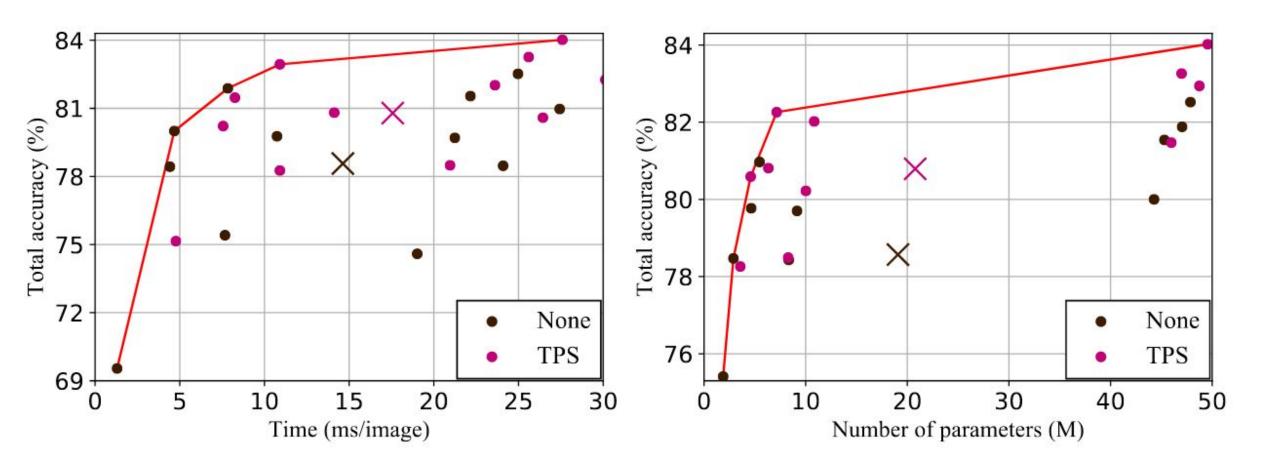


#### Possible Architecture



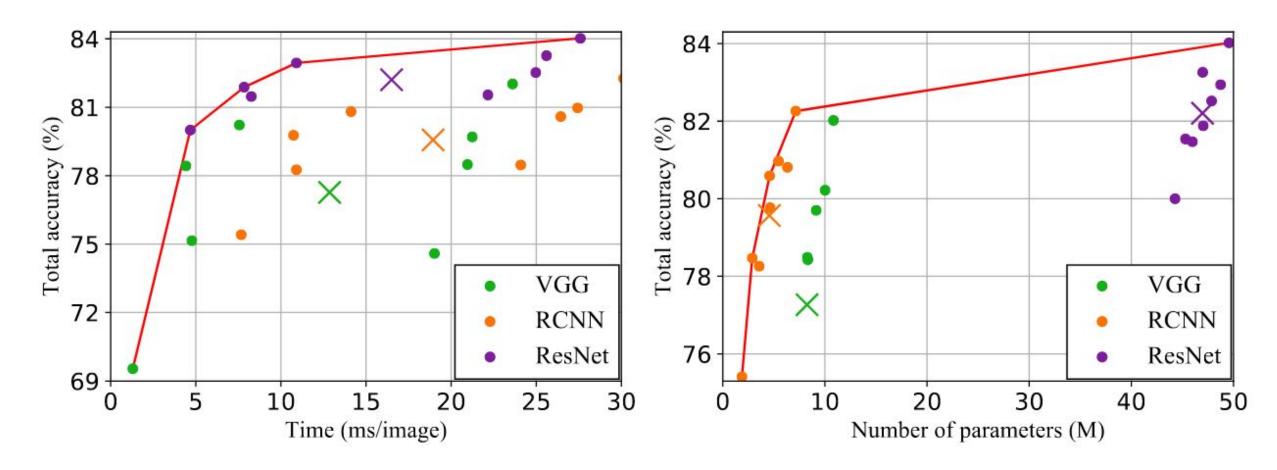


#### TPS / non TPS comparison



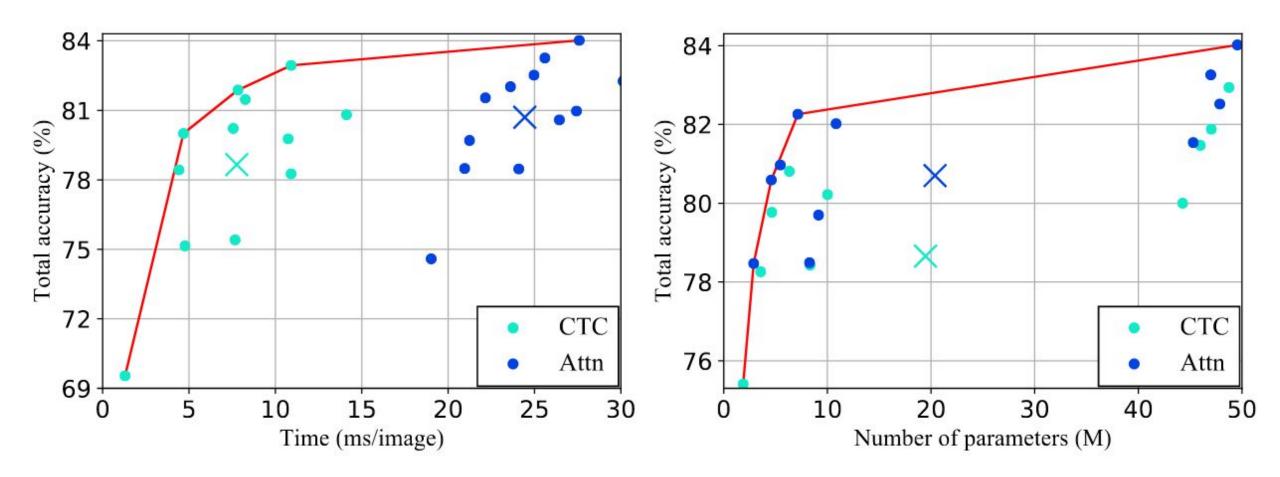


#### Feature extractor comparison





## Prediction step comparison





# **Questions?**



## **Clovai results**

TPS	VGG	None	CTC	80.0	78.0	90.1	89.7	88.7	87.5	65.1	60.6	65.5	57.0	75.1	4.8	7.3
			Attn	82.9	82.3	92.0	91.7	90.5	89.2	69.4	64.2	73.0	62.2	78.5	21.0	8.3
		BiLSTM	CTC	84.6	83.8	93.3	92.9	91.2	89.4	72.4	66.8	74.0	66.8	80.2	7.6	10.0
			Attn	86.2	85.8	93.9	93.7	92.6	91.1	74.5	68.9	76.2	70.4	82.0	23.6	10.8
	RCNN	None	CTC	82.8	81.7	92.0	91.6	89.5	88.4	69.8	64.6	71.3	61.2	78.3	10.9	3.6
			Attn	85.1	84.0	93.1	93.1	91.5	90.2	72.4	66.8	75.6	64.9	80.6	26.4	4.6
		BiLSTM	CTC	85.1	84.3	93.5	93.1	91.4	89.6	73.4	67.7	74.4	69.1	80.8	14.1	6.3
			Attn	86.3	85.7	94.0	94.0	92.8	91.1	75.0	69.2	77.7	70.1	82.3	30.1	7.2
	ResNet	None	CTC	85.0	85.7	94.0	93.6	92.5	90.8	74.6	68.8	75.2	71.0	81.5	8.3	46.0
			Attn	87.1	87.1	94.3	93.9	93.2	91.8	76.5	70.6	78.9	73.2	83.3	25.6	47.0
		BiLSTM	CTC	87.0	86.9	94.4	94.0	92.8	91.5	76.1	70.3	77.5	71.7	82.9	10.9	48.7
			Attn	87.9	87.5	94.9	94.4	93.6	92.3	77.6	71.8	79.2	74.0	84.0	27.6	49.6

