# DETECTION OF THE TYPES OF IMAGES (PHOTOS, VIDEOS) PUBLISHED BY POLITICAL PAGES ON FACEBOOK

(Détection des types d'images (photos, vidéos) publiés par les pages politiques de Facebook)

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#### INTRODUCTION



# LA RÉPUBLIQUE En Marche!









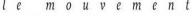






Les CENTRISTES





# **OBJECTIVES**

- To detect object in an image published by political parties or by the media.
- Compare different object detecting tools and state-of-the-art neural network models and their respective annotating methods.

### OBJECT DETECTION

- 1. To enable analysis on political marketing with political objects such as.
  - Political Symbols
  - ➤ Meeting Context





## OBJECT DETECTION

2. To detect types of images published by political parties to promote their candidates

3. To identify and quantify the main subjects of the images shared by the political communities.

#### PREVIOUS RESEARCH

'A quick glance at an image is sufficient for a human to point out and describe an immense amount of details about the visual scene. However, this remarkable ability has proven to be an elusive task for our visual recognition models' — (Karpathy & Fei-Fei, 2017).

'Automatically describing the content of an image is a fundamental problem in artificial intelligence that connects computer vision and natural language processing.'

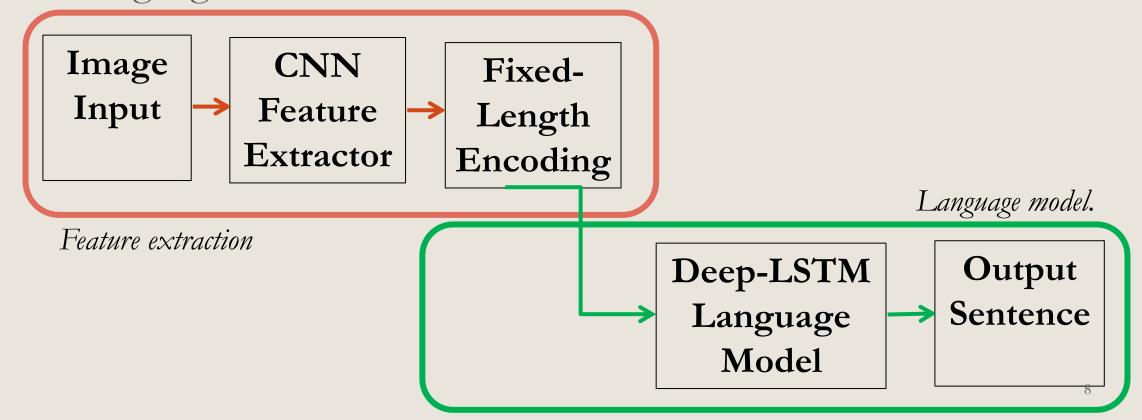
— (Vinyals, et al., 2014).

#### PREVIOUS RESEARCH

- Template-based methods
- Nearest neighbor methods
- Modifying existing captions
- Neural network models

#### NEURAL NETWORK MODELS

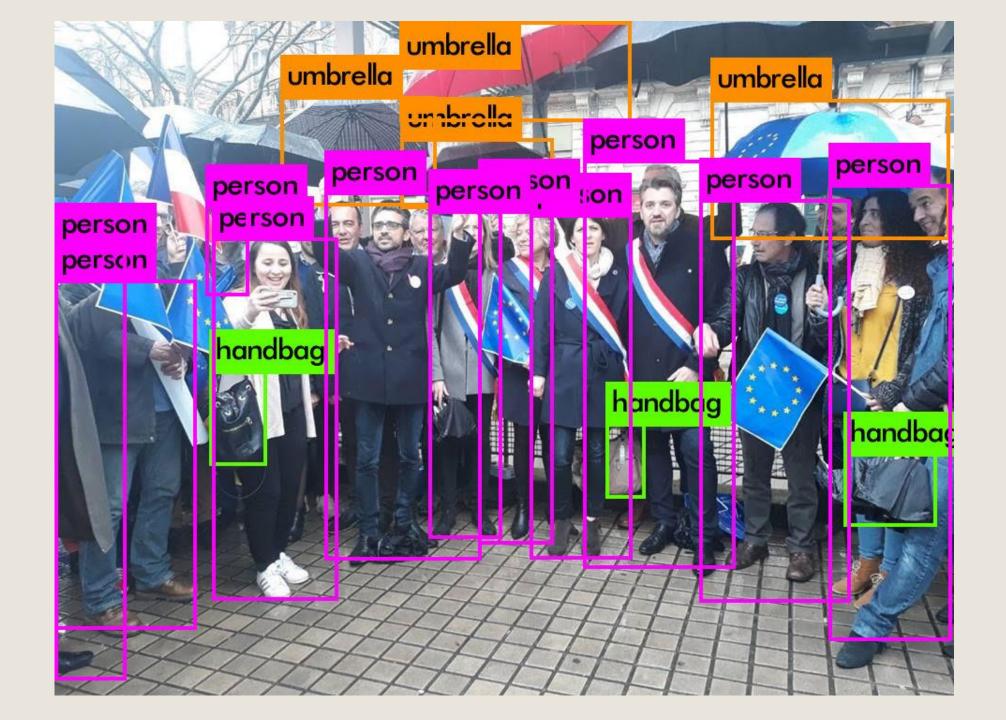
- Feature extraction
- Language model.



# AUTOMATIC IMAGE ANNOTATION

• Object detection and localization:





## AUTOMATIC IMAGE ANNOTATION

- Object detection using CNN
  - ➤ One stage
  - > Two stage
- Localization
  - ➤ Bounding box
  - >Image segmentation

#### SOA CNN MODELS

- RetinaNet
- Deformable Part Models v5
- Region-CNN
- YOLO, YOLO9000, TinyYOLO, YOLOv3
- TensorFlow by Google AI
- Fast-CNN
- Faster R-CNN +++
- Faster R-CNN by G-RMI
- Faster R-CNN with FPN (feature pyramid network)

#### ANNOTATION TOOLS

- Anno-mage
- LabelIMG
- LabelBox
- Supervise.ly
- Etc.

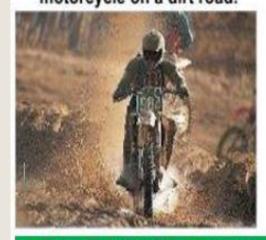
#### **COMMON METRICS**

- Mean average precision (mAP).
- Frame per second (FPS)
- Speed/Duration
- Multiscale prediction
- Backbone classifier

# ANNOTATION TYPES WITH DEEP LEARNING TOOLS

1. Context-based Textual descriptions/Caption

A person riding a motorcycle on a dirt road.



Describes without errors

Two dogs play in the grass.

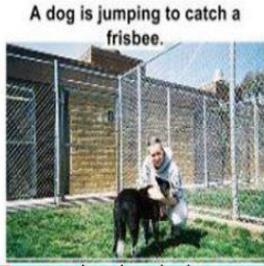


Describes with minor

A skateboarder does a trick



Somewhat related to the image



Unrelated to the image

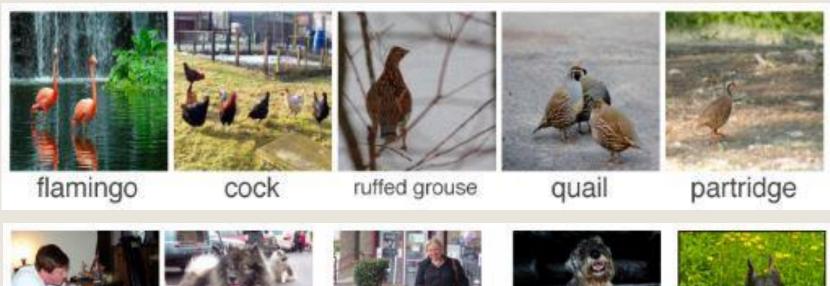
# ANNOTATION TYPES WITH DEEP LEARNING TOOLS

2. Descriptions of image regions



# ANNOTATION TYPES WITH DEEP LEARNING TOOLS

3. Description/classification of image into known class













#### CONCLUSION

Different models will be identified and their annotating strengths and cons will be documented.

To identify the main subjects of the photos/images in order to compare them with the topics of Facebook posts that the lexicometric analysis (the topics model) has been identify.

(This analysis is to be done by Julien Figeac and Pierre Ratinaud.)

#### **CONCLUSION**

Still not easy-peasy for computers!



sharpei or croissant?

#### **CONCLUSION**

Still not easy-peasy for computers!



teddy bear or cute dog

## OPEN QUESTIONS

- Can we identify the topics of the images?: Social movements, riots, strikes
  - Terrorist events
  - Economic events (taxes, fraud scandal, etc.)
- Can we analyse the images when it centered on a person (top of the body, face)?
  - Body language?
  - -Gender?
- Public gatherings and events?
  - Important public events (e.g. "Agricultural exhibition in Paris" and other public events where politicians are present)

#### REFERENCES

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