American Sign Language Recognition using Convolutional Neural Networks

Fatima-Zahrae El-Qoraychy, Yazan Mualla

CIAD, University of Technology of Belfort-Montbéliard (UTBM)

fatima.el-qoraychy@utbm.fr yazan.mualla@utbm.fr

Conference on Advances in Human-oriented and Personalized Mechanisms, **Technologies, and Services CENTRIC 2023** Valencia, Spain, November 13-17, 2023





Name: Fatima-Zahrae El-Qoraychy

- Degree: Master's in Data Science and Artificial Intelligence
- University: University of Sorbonne Paris Nord, France
- □ Year of Degree: 2023
- Doctoral Studies: Currently pursuing a Ph.D. at the CIAD Laboratory, University of UBFC
- Research Area: Socially Smart Connected and Autonomous Vehicles
- □ Research Focus: Enhancing the safety and efficiency of cooperative intersections using Deep Reinforcement Learning











Problematic

UBFC

RÉPUBLIQUE FRANÇAISE

Égalité Fraternite

How can we use artificial intelligence to predict and interpret hand gestures in sign language to facilitate communication and interaction between deaf and hard-ofhearing individuals and their environment?

utbm



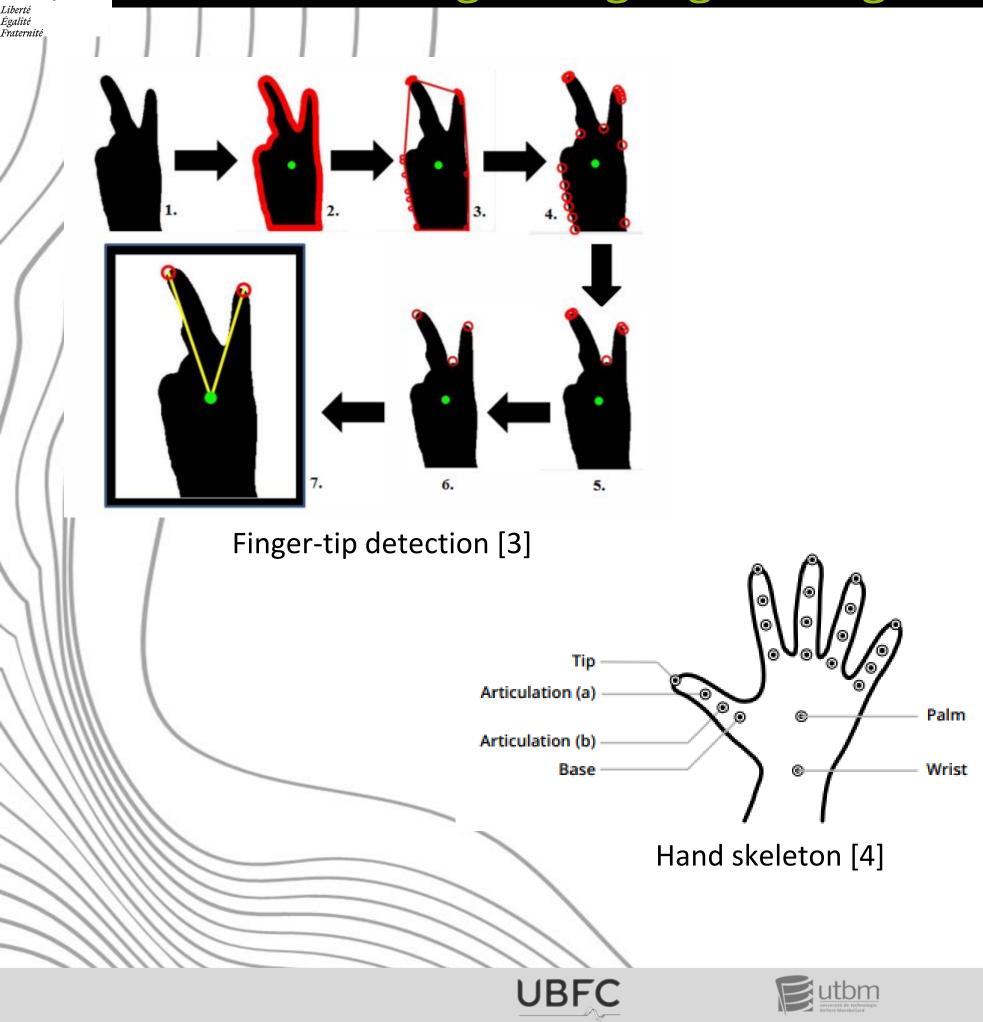








RÉPUBLIQUE FRANÇAISE Advances in Sign Language Recognition



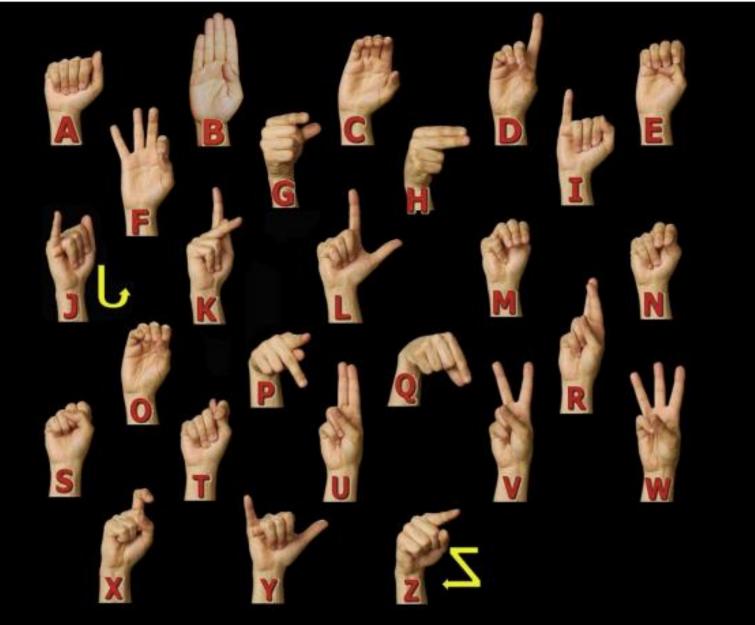






ARNO

ARTS



American sign language finger spelling alphabet [5]





RÉPUBLIQUE FRANÇAISE **Existing Solution**

e

DAMION JOYNER · COPIED FROM BARRY +352,-100 · 1Y AGO · 432 VIEWS

Link [6]

Égalité Fraternité

Sign-Language-Classification-CNN-VGG19

Python · VGG-19, ASL Fingerspelling Images (RGB & Depth), [Private Datasource]





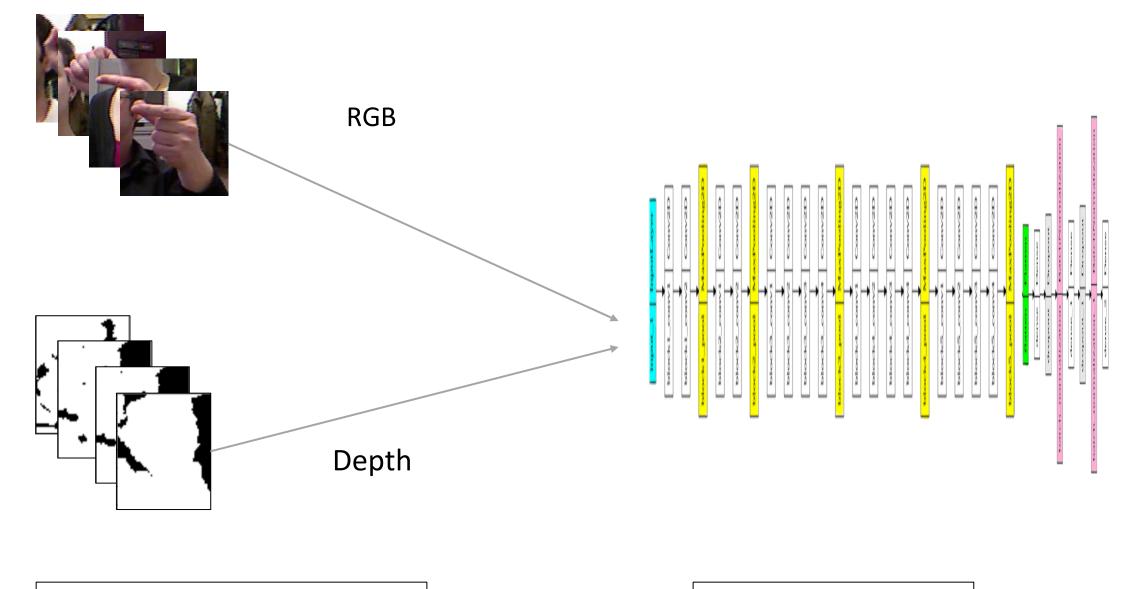












100 000 Labeled images [7]

Classification model











Prediction



Existing Method Performance Analysis

- (B) 🎍

Égalité Fraternité

RÉPUBLIQUE FRANCAISE

Confusion matrices

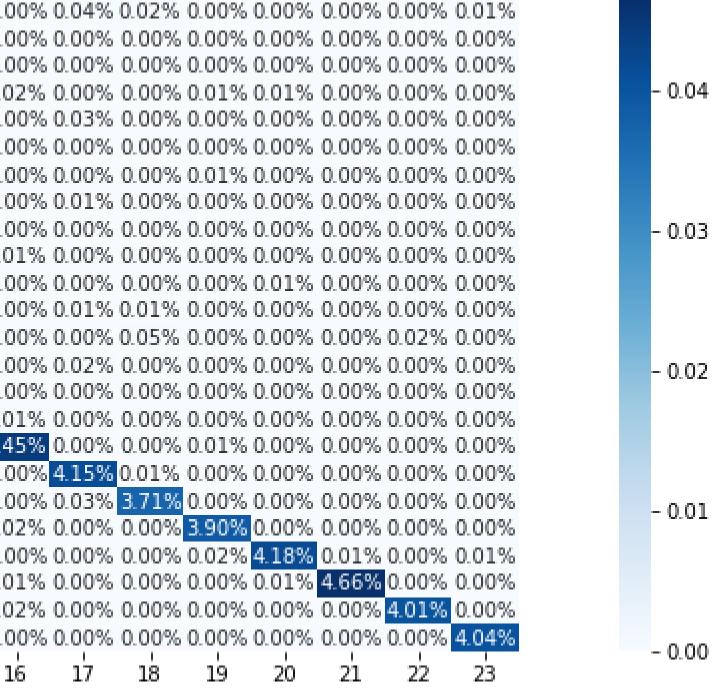
.18% 0.00% 0.00% 0.00% 0.01% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.04% 0.02% 0.00% 0.00% 0.00% 0.00% 0.01% -0.00% 4.14% 0.01% 0.00% 0.00% 0.01% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 4.52% 0.00% 0.01% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 0.01% 0.01% 0.00%-0.00% 0.00% 0.01% 4.22% 0.00% 0.00% 0.01% 0.00% 0.00% 0.00% 0.01% 0.00% 0.01% 0.01% 0.01% 0.02% 0.00% 0.00% 0.00% 0.01% 0.00% 0.00% 0.00% → -0.01% 0.01% 0.03% 0.01% 4.06% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 0.00% 0.00% 0.00% 0.03% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% -0.00% 0.01% 0.00% -0.01% 0.01% 0.00% 0.00% 0.00% 0.00% 0.00% 0.04% 0.00% 0.01% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% -0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.05% 4.22% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% ∞ -0.00% -0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 0.00% 0.00% 0.00% 0.00% 0.01% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% -0.00% 10 -0.01% 0.00% 0.01% 0.01% 0.00% 0.00% 0.01% 0.01% 0.00% 0.00% 0.00% 0.02% 0.02% 0.02% 0.00% 0.00% 0.00% 0.01% 0.01% 0.00% 0.00% 0.00% 0.00% 0.00% -0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 0.00% 0.02% 4.00% 0.02% 0.00% 0.00% 0.00% 0.00% 0.05% 0.00% 0.00% 0.00% 0.02% 0.00% <u>evi</u> -0.00% 0.00% 0.00% 0.02% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.02% 0.02% 3.81% 0.01% 0.00% 0.00% 0.02% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% В -0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 0.00% 0.00% <u>3.95%</u> 0.02% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% <u>चि</u>र -0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.02% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 4.16% 0.01% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% LO. -0.00% ശ -0.03% 0.00% 0.00% 0.00% 0.02% 0.00% 0.00% 0.00% 0.00% 0.00% 0.06% 0.00% 0.03% 0.00% 0.00% 0.00% 0.00% 0.01% 0.00% 0.00% 0.00% 0.00% 0.00% -0.03% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 0.03% 0.00% 0.00% 0.00% 0.00% 0.03% 3.71% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 13-0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.02% 0.00% 0.00% <u>3.90%</u> 0.00% 0.00% 0.00% 0.00% σ -0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 0.02% 0.00% 0.01% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.02% 4.18% 0.01% 0.00% 0.01% 2 -0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 0.00% 0.00% 0.00% 0.01% 4.66% 0.00% 0.00% ನ -0.00% 0.00% 0.00% 0.02% 0.01% 0.00% 0.00% 0.00% 0.00% 0.01% 0.00% 0.01% 0.01% 0.01% 0.00% 0.02% 0.00% 0.00% 0.00% 0.00% 0.00% 4.01% 0.00% -0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 0.00% 9. 100 8 11 12 13 14 15









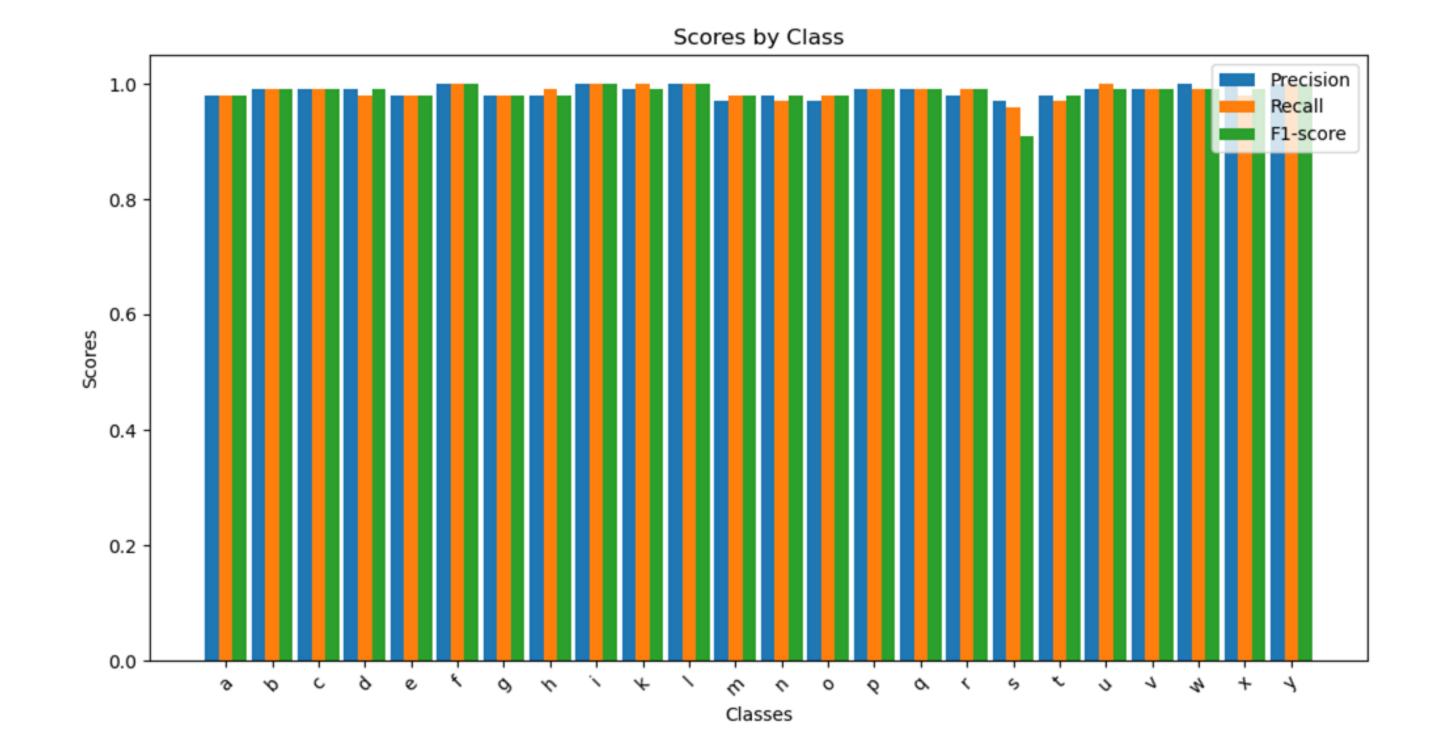




Existing Method Performance Analysis

RÉPUBLIQUE FRANÇAISE

Liberté Égalité Fraternité





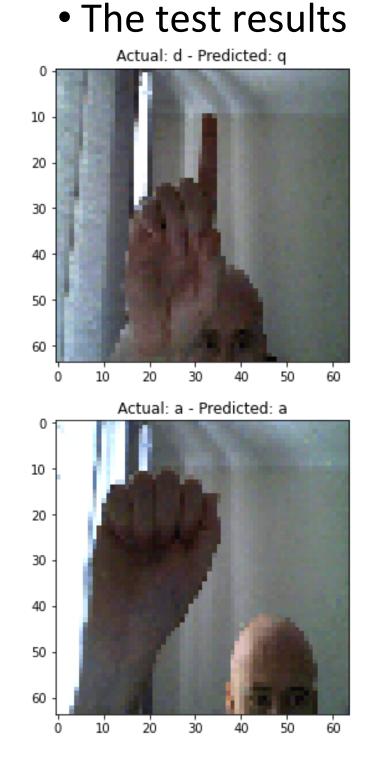






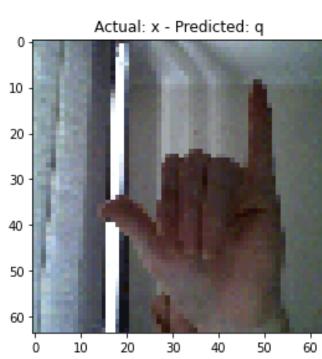


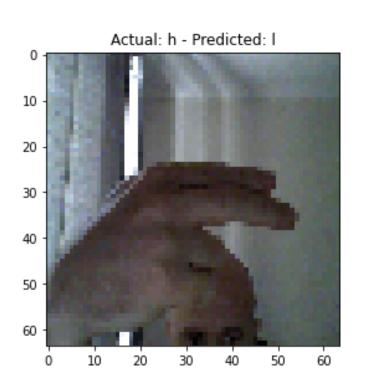
Existing Method Test Results



RÉPUBLIQUE FRANÇAISE

Liberté Égalité Fraternité



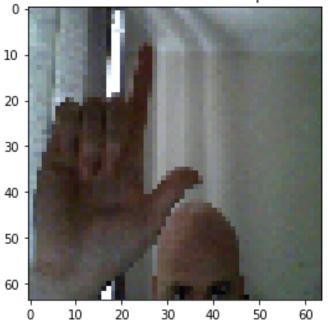




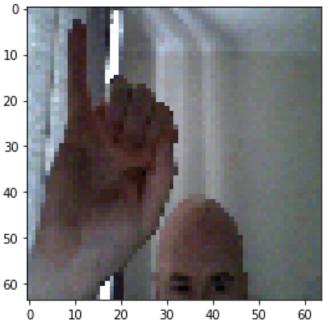








Actual: k - Predicted: q



Actual: i - Predicted: c





Limitations of the Classification Model

- Lack of Data Diversity
- **Overfitting to Specific Hand Features**

Our Proposed Enhancements

- **Data Diversification**
- Data Augmentation
- Using Segmented Images







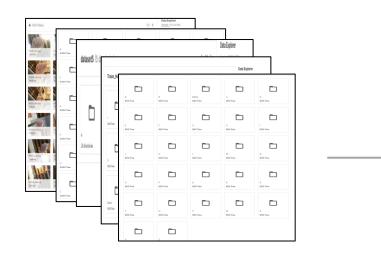








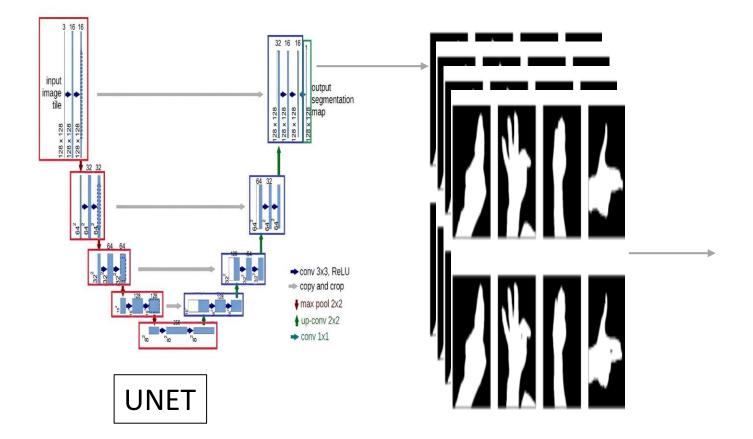
Proposed Solution: Structure





	directories	labels	images
0	А	а	A/a/color_0_0002.png
1	А	а	A/a/color_0_0003.png
2	А	а	A/a/color_0_0004.png
3	А	а	A/a/color_0_0005.png
4	А	а	A/a/color_0_0006.png

436,433 Labeled images							
[7] [8] [9] [10] [11]							



	directories	labels	images
0	А	а	A/a/color_0_0002_mask.png
1	А	а	A/a/color_0_0003_mask.png
2	А	а	A/a/color_0_0004_mask.png
3	А	а	A/a/color_0_0005_mask.png
4	А	а	A/a/color_0_0006_mask.png

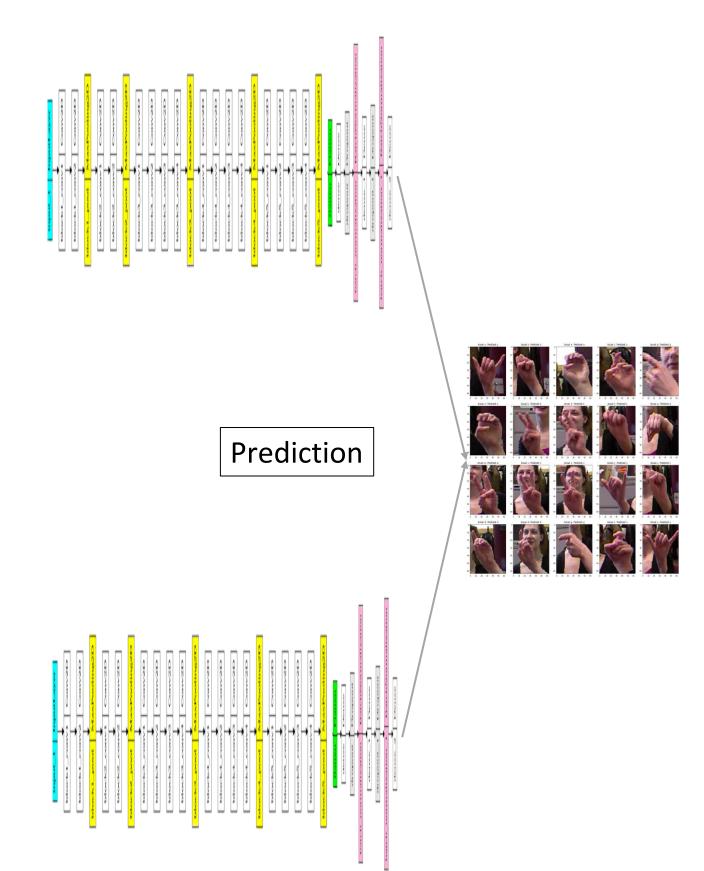


Université de technologie









RÉPUBLIQUE FRANÇAISE Égalité Fraternite

Proposed Solution: Performance Analysis

Confusion matrices

1 -0.03% 4.07% 0.01% 0.01% 0.01% 0.01% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.02% 0.01% 0.00% 0.01% 0.00% 0.02% 0.00 6-0.01%0.00%0.00%0.00%0.00%0.00%<u>4.24%</u>0.09%0.00%0.00%0.01%0.00%0.00%0.00%0.01%0.00%0 15 - 0.00%16 - 0.00% 0.00% 0.00% 0.00% 0.00% 0.00% 0.01% 0.01% 0.00%14 15 16 10 11 12 13 17 0 З. 5 8 9 2 6 7









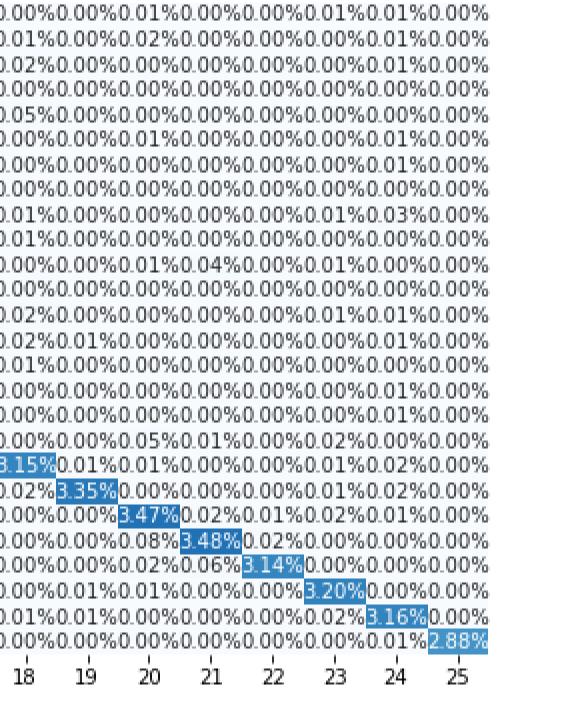
- 0.04

- 0.03

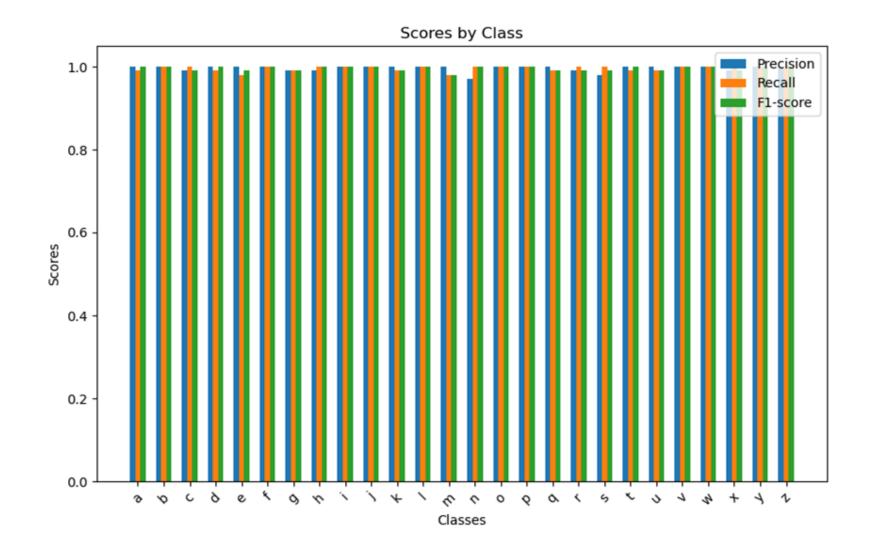
- 0.02

- 0.01

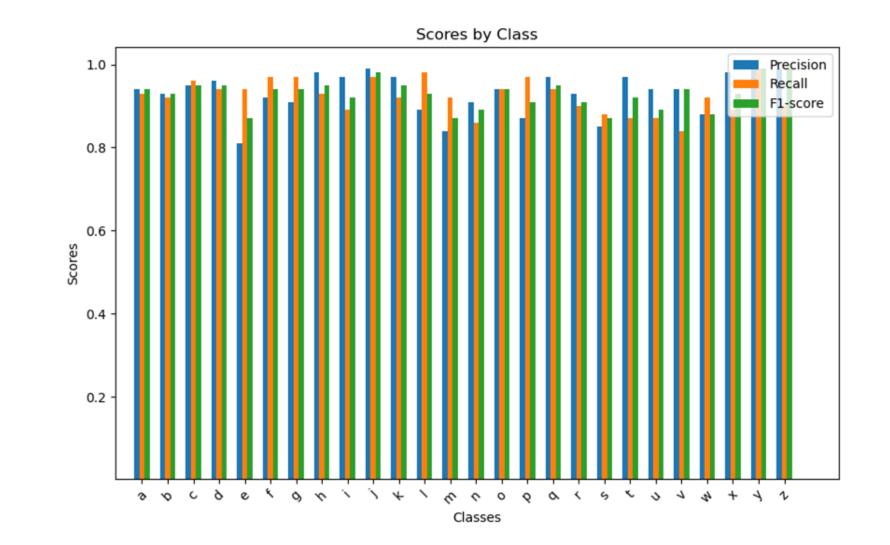
- 0.00







Liberté Égalité Fraternité



Results of the RGB model









Results of the mask model





RÉPUBLIQUE FRANÇAISE Liberté Égalité Fraternité

Weighted average	Accuracy	F1 score
RGB model	0.99	0.99
Mask model	0,93	0,93









Precision

0.99

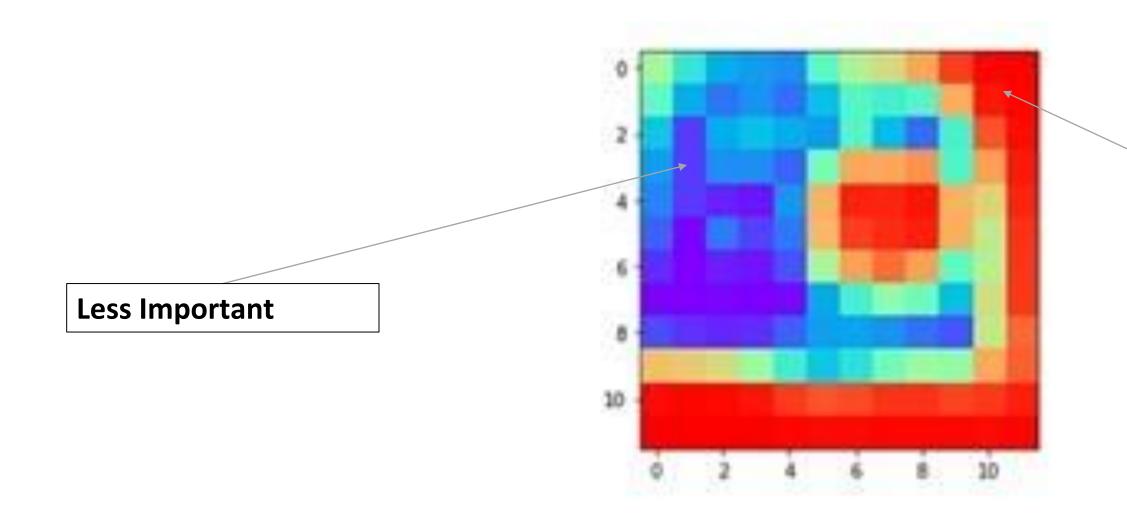
0,93



Enhancing Model Explainability with Grad-CAM

Égalité Fraternit

Grad-CAM (Gradient-weighted Class Activation Mapping):
An interpretive artificial intelligence technique used to identify and illustrate noteworthy regions within an image during the classification process
It generates a heat map that emphasizes the areas most impactful for predicting the class











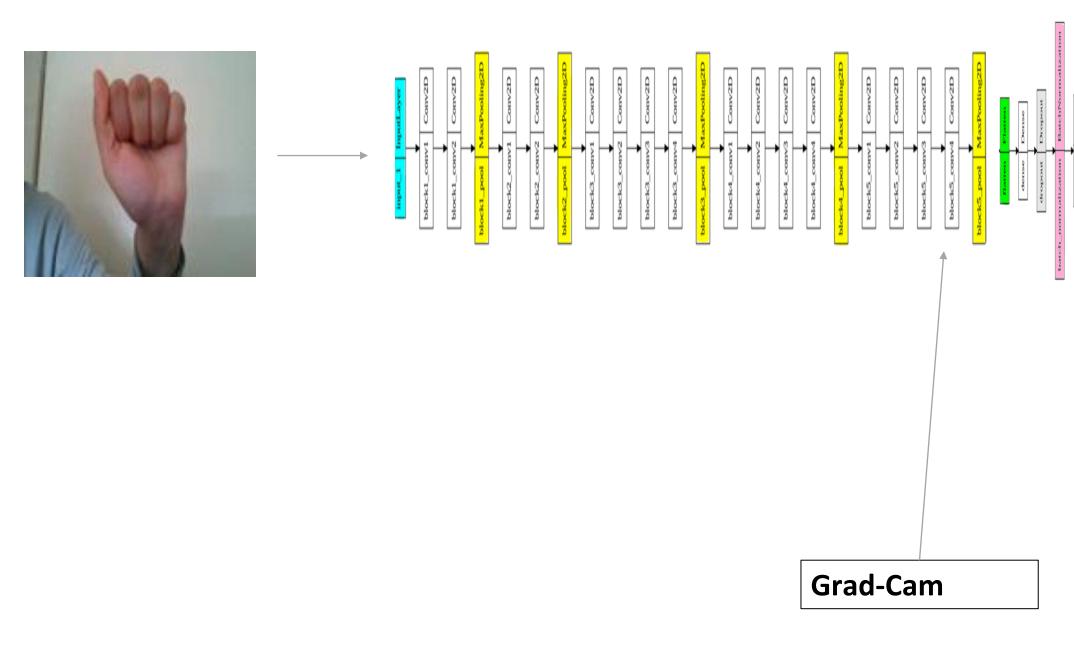
More Important





Enhancing Model Explainability with Grad-CAM

Grad-Cam on RGB Image Classification Model







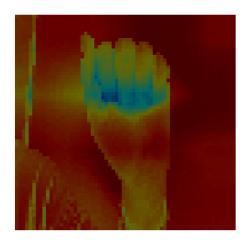








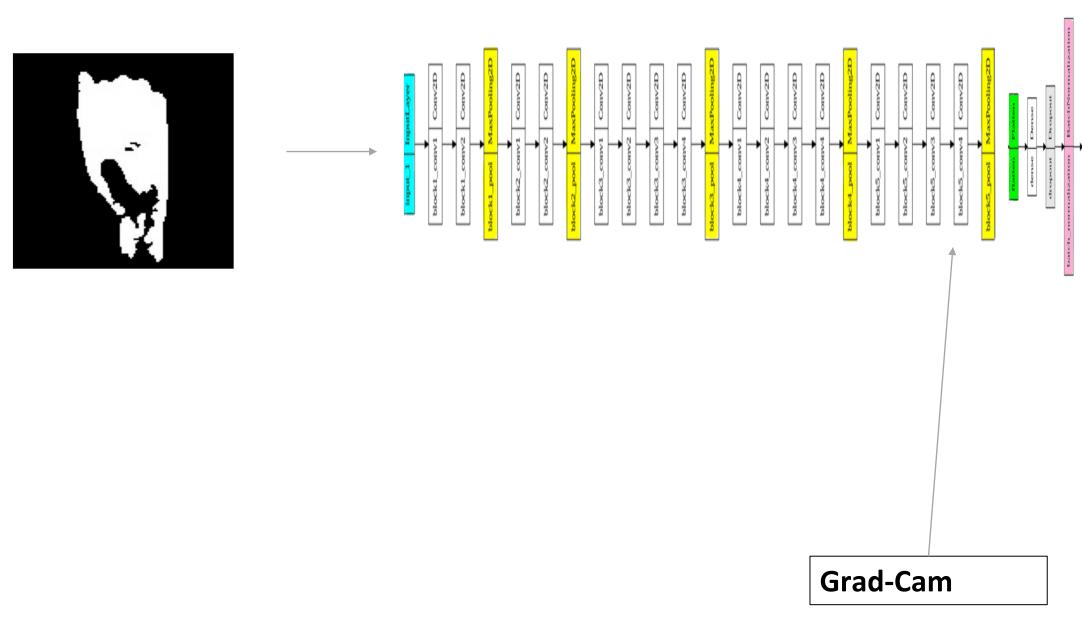
Prediction a







Grad-Cam on Mask Image Classification Model







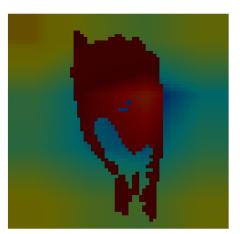








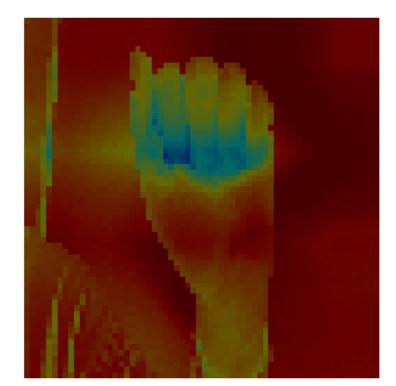
Prediction a







Enhancing Model Explainability with Grad-CAM



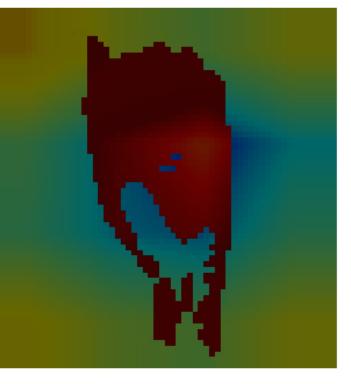
RGB model











Mask model

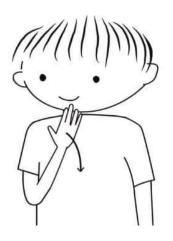




- Extend the project to handle expressions
- Extend the work to process video and enable real-time communication with a hard of hearing or deaf person
- Development of a framework that will be used to communicate with this category of individuals
- Utilize other explainable models















THANK YOU





- References
- [1] https://www.ciad-lab.fr/yazan_mualla/
- [2] https://www.ciad-lab.fr/fatima-zahrae_el-goraychy/
- Anup Kumar, Karun Thankachan, and Mevin M. Dominic. Sign language recognition. In 2016 3rd International Conference on [3] Recent Advances in Information Technology (RAIT), pages 422-428, 2016.
- [4] Guillaume Devineau, Wang Xi, Fabien Moutarde, and Jie Yang. Dee learning for hand gesture recognition on skeletal data. In 2018 13th IEEE International Conference on Automatic Face Gesture Recognition (FG 2018), page 106–113. IEEE, 2018
- [5] Brandon Garcia and Sigberto Alarcon Viesca. Real-time american sign language recognition with convolutional neural networks. volume 2, page 8. Stanford University Stanford, CA, USA, 2016.
- [6] https://www.kaggle.com/code/damionjoyner/sign-language-classification-cnn-vgg19
- [7] https://www.kaggle.com/datasets/mrgeislinger/asl-rgb-depth-fingerspelling-spelling-it-out
- [8] https://www.kaggle.com/datasets/grassknoted/asl-alphabet
- [9] https://www.kaggle.com/datasets/danrasband/asl-alphabet-test
- [10] https://www.kaggle.com/datasets/debashishsau/aslamerican-sign-language-aplhabet-dataset
- [11] https://www.kaggle.com/datasets/lexset/synthetic-asl-alphabet









