

### The Key Points: Using feature importance to identify shortcomings in sign language recognition models

RUTH HOLMES, ELLEN RUSHE, ANTHONY VENTRESQUE



**Trinity College Dublin** Coláiste na Tríonóide, Baile Átha Cliath ne University of Dublin









lorizon 2020 **European Union Funding Research & Innovation** 

### **AUTHORS**

COMPLEX SOFTWARE LAB - TRINITY COLLEGE DUBLIN



RUTH HOLMES

holmesru@tcd.ie



ELLEN RUSHE

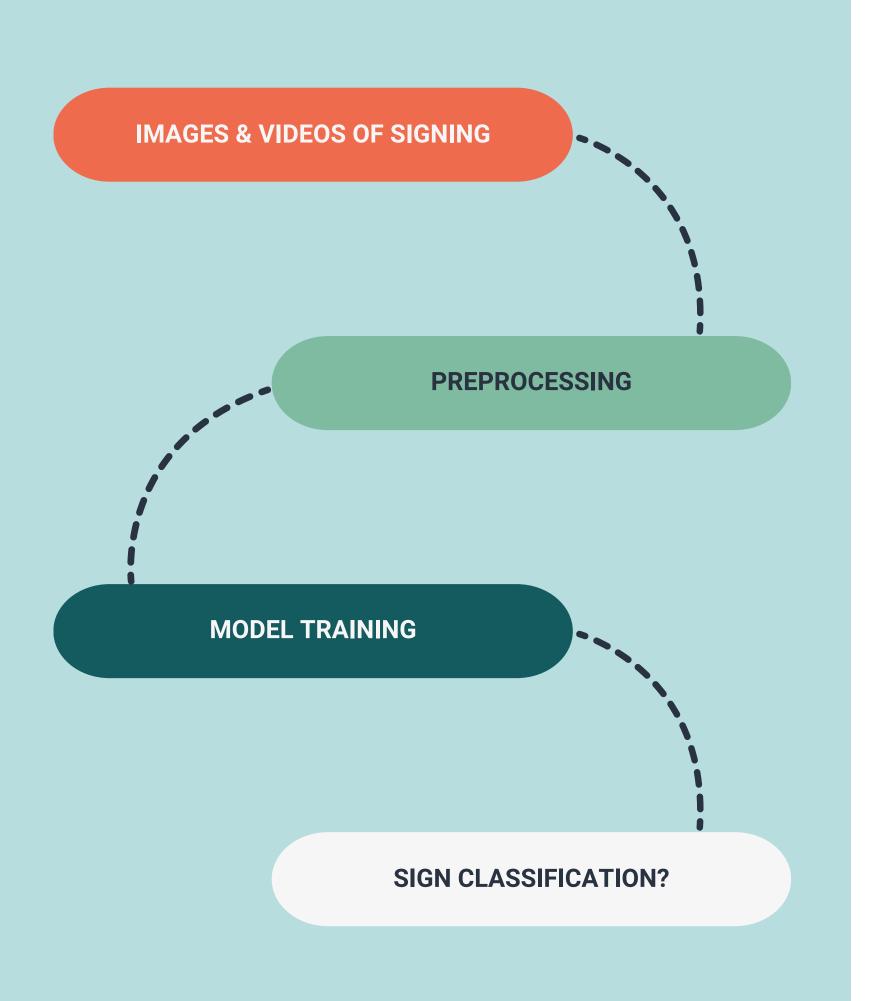
ellen.rushe@tcd.ie

02



#### ANTHONY VENTRESQUE

anthony.ventresque@tcd.ie



# SIGN LANGUAGE RECOGNITION

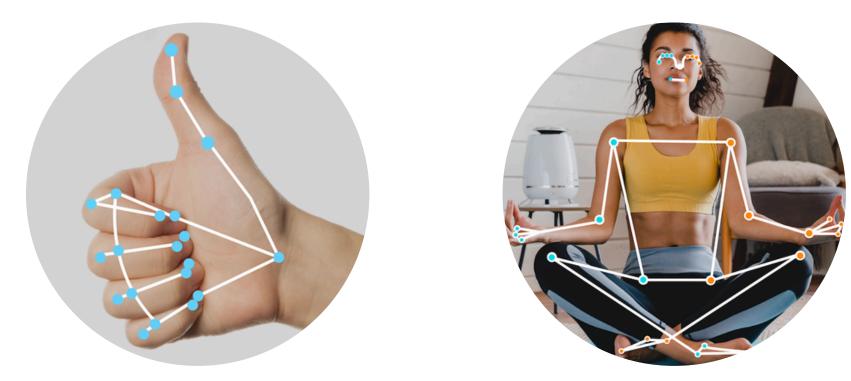
In the field of Computational Linguistics, Sign Language Recognition (SLR) sits at the intersection of Computer Vision and Natural Language Processing.

#### **DATA-BASED CHALLENGES:**

- Complex visual modality
- Scarcity
- Lack of diversity

# **POSE-ESTIMATION TOOLS**

Reliable means of detecting the salient parts of the body - termed *keypoints*.



Google. (2023). Hand gesture. MediaPipe | Google for Developers. https://developers.google.com/mediapipe Google. (2023). *Pose detector*. MediaPipe | Google for Developers. https://developers.google.com/mediapipe

#### DIMENSIONALITY

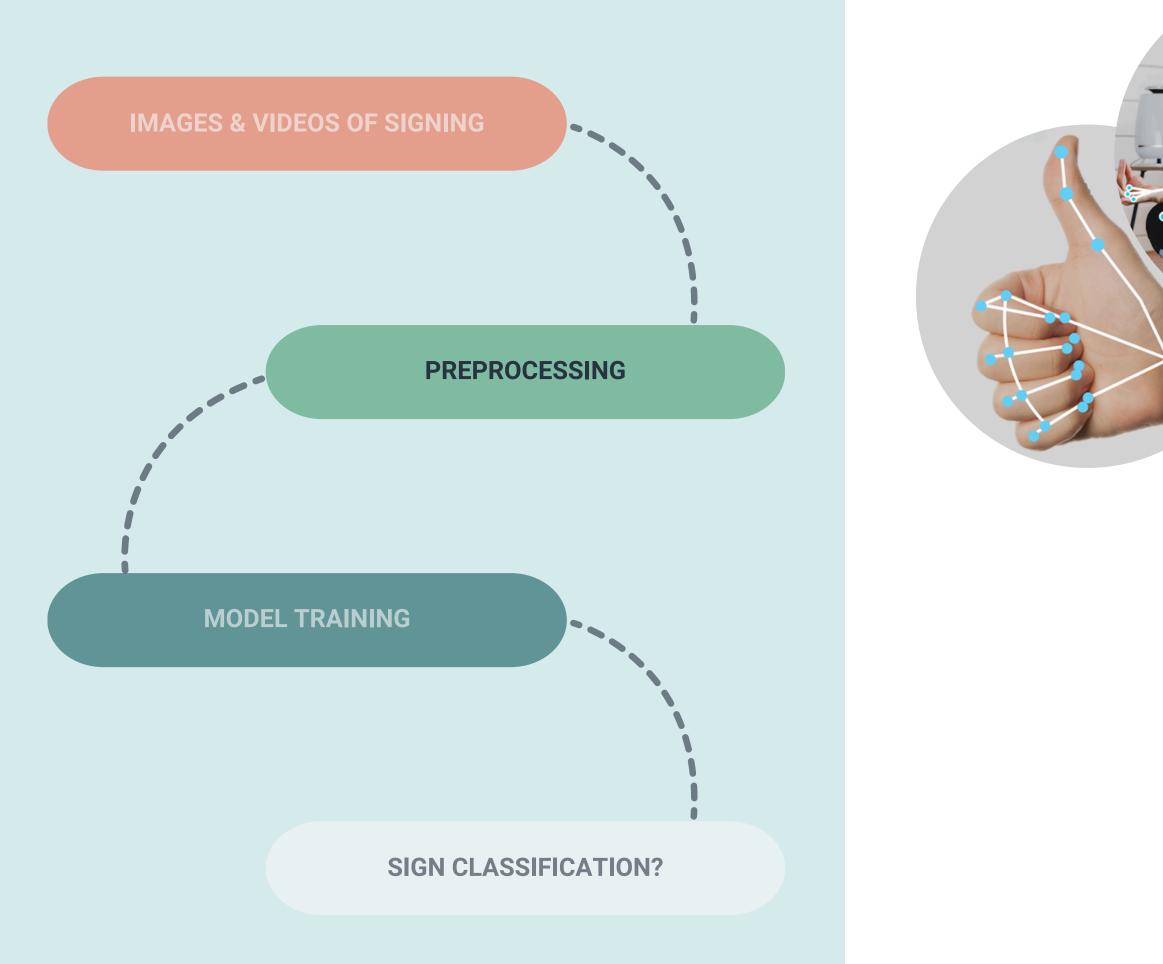
Keypoints offer a structured input representation and a significantly reduced dimensionality compared to visual data.

Extensive training and diverse data enhances our ability to cope with different visual conditions such as lighting and background scenery.

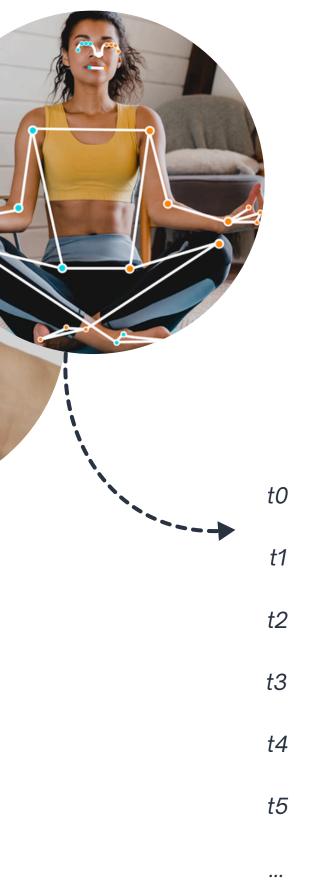
#### **HUMAN VARIATION**

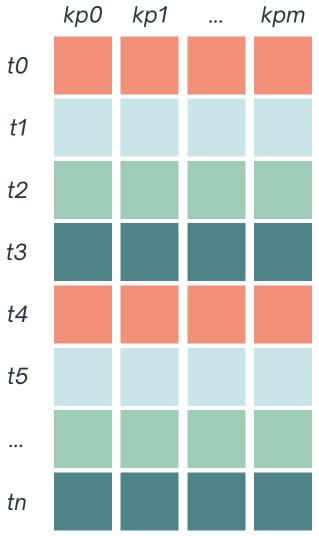
This also better equips us for signer variation and limits the personal characteristics learnable by subsequent SLR models.

#### **VISUAL CONDITIONS**



Google. (2023). *Hand gesture*. MediaPipe | Google for Developers. https://developers.google.com/mediapipe Google. (2023). *Pose detector*. MediaPipe | Google for Developers. https://developers.google.com/mediapipe





## **DATA - CORPUS VGT**

#### FLEMISH SIGN LANGUAGE (VLAAMSE GEBARENTAAL, VGT).

Continuous signing videos processed into clips based on available gloss-tier annotations.

Stratified split to ensure similar class distribution in all subsets.

Grouped split to ensure signer-independent data configuration.

TRAINING	VAI
88 participants	12 p
19,267 samples	2,70
292 classes	29

Mieke Van Herreweghe, Myriam Vermeerbergen, Eline Demey, Hannes De Durpel, Hilde Nyffels, and Sam Verstraete. 2015. Het Corpus VGT. Een digitaal open access corpus van videos and annotaties van Vlaamse Gebarentaal, ontwikkeldaan de Universiteit Gent ism KU Leuven. www.corpusvgt.be.

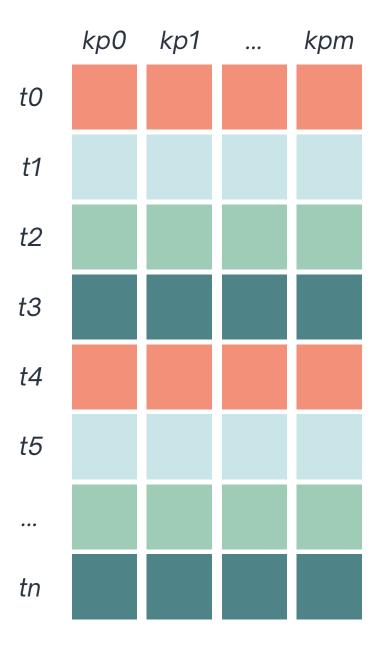


#### LIDATION

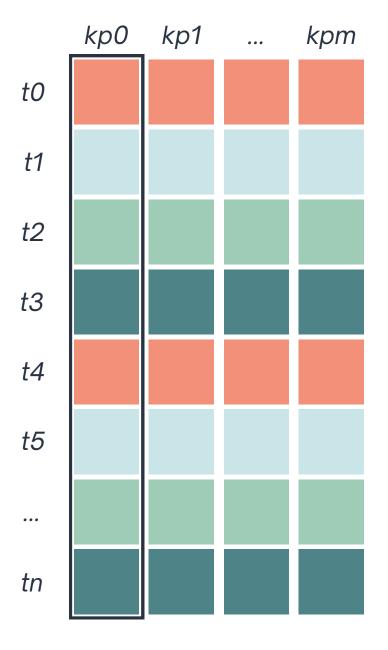
participants 02 samples 92 classes

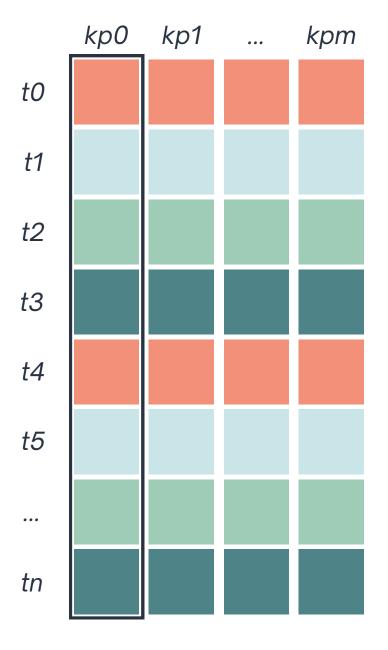
#### TEST

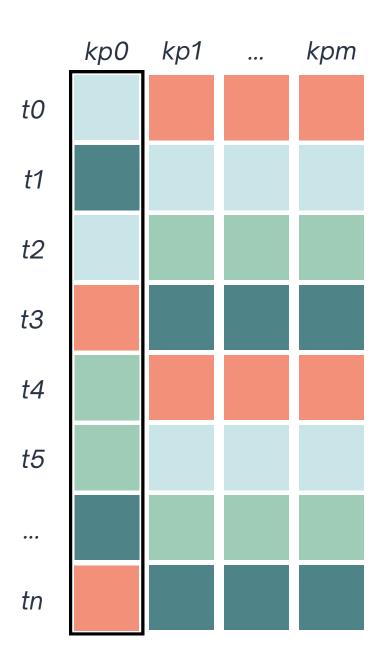
11 participants2,998 samples292 classes



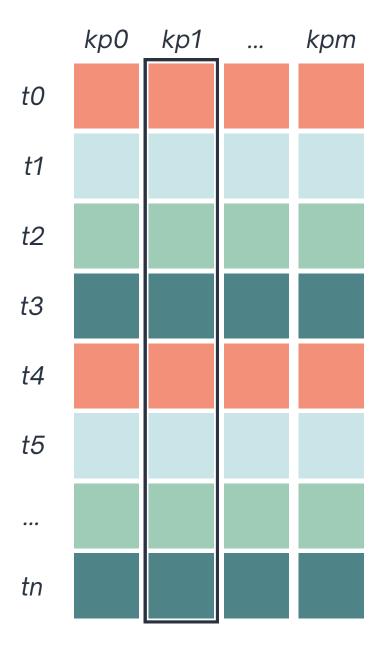


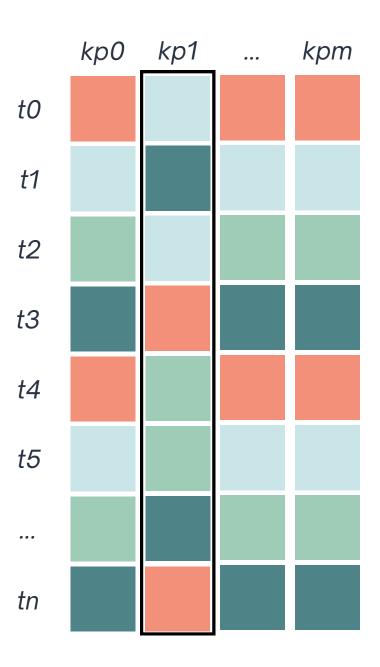




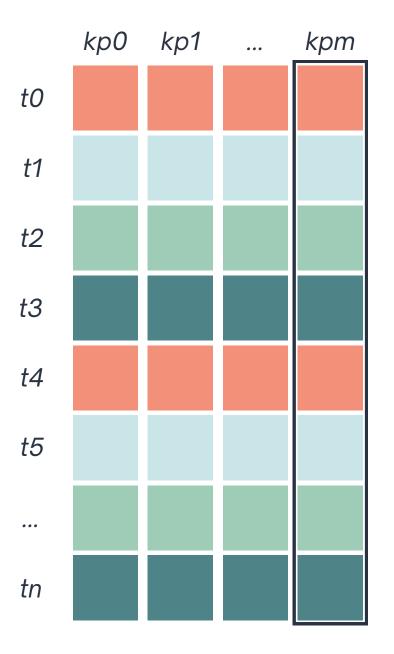




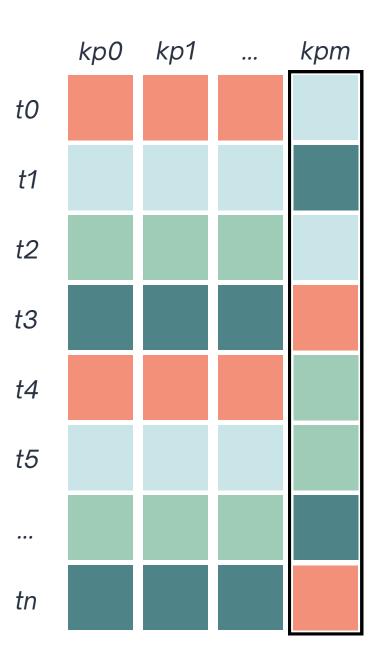




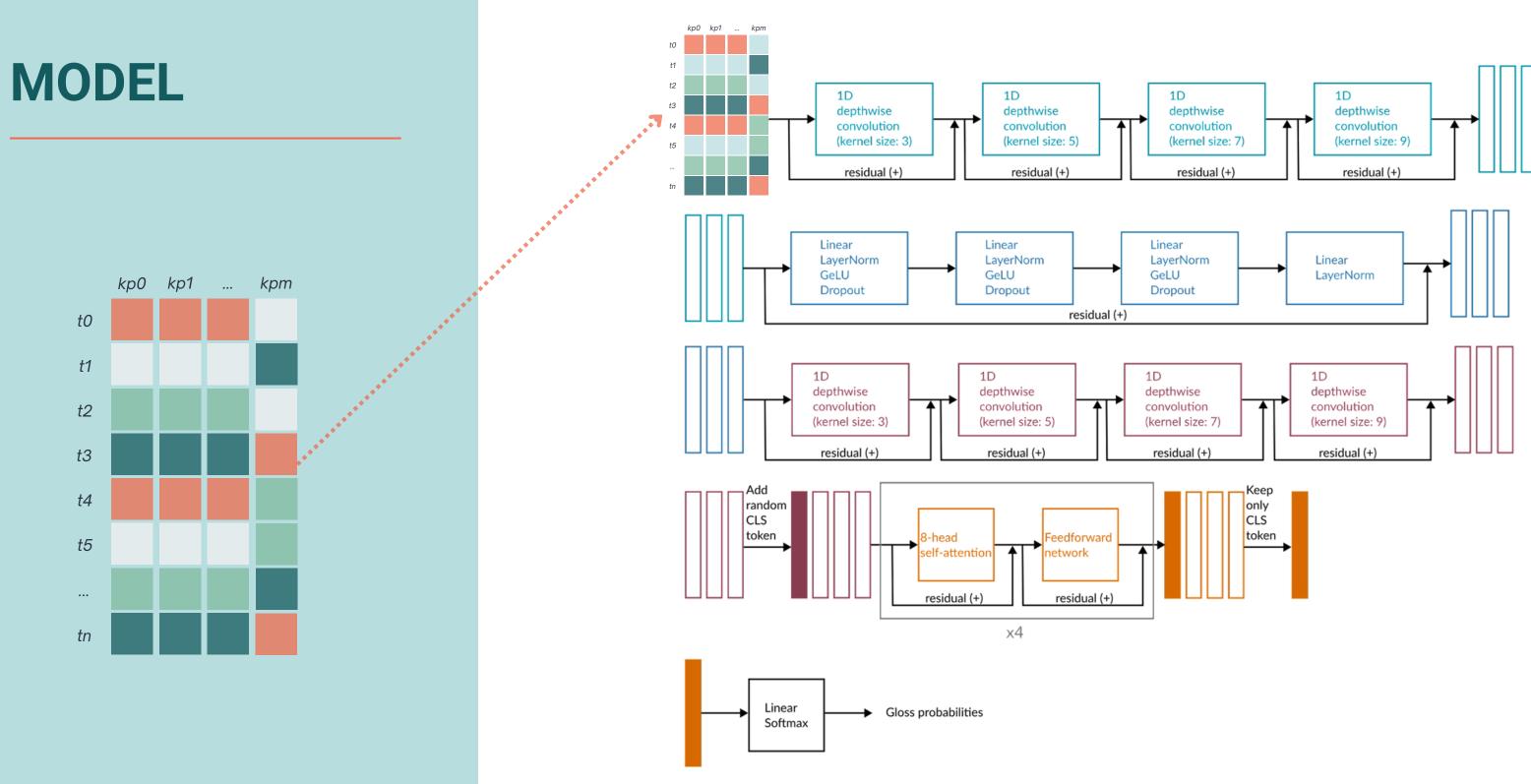




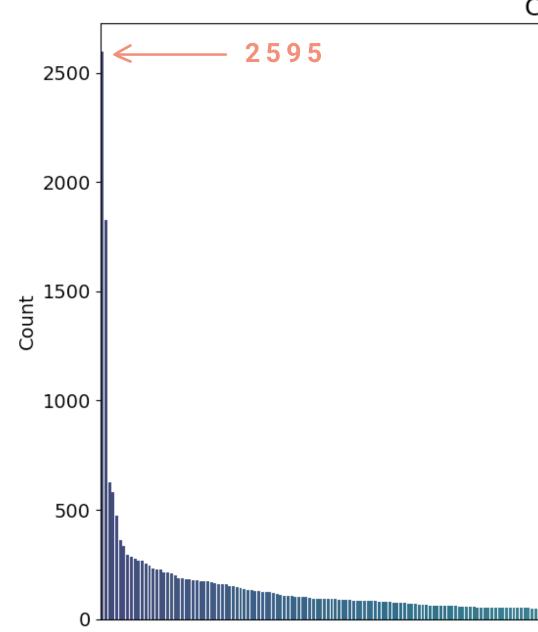








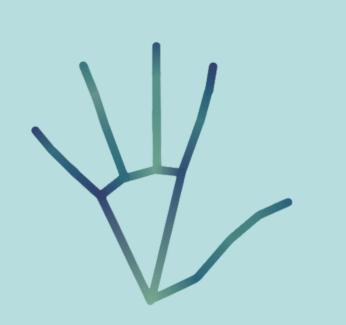
### **CLASS IMBALANCE**

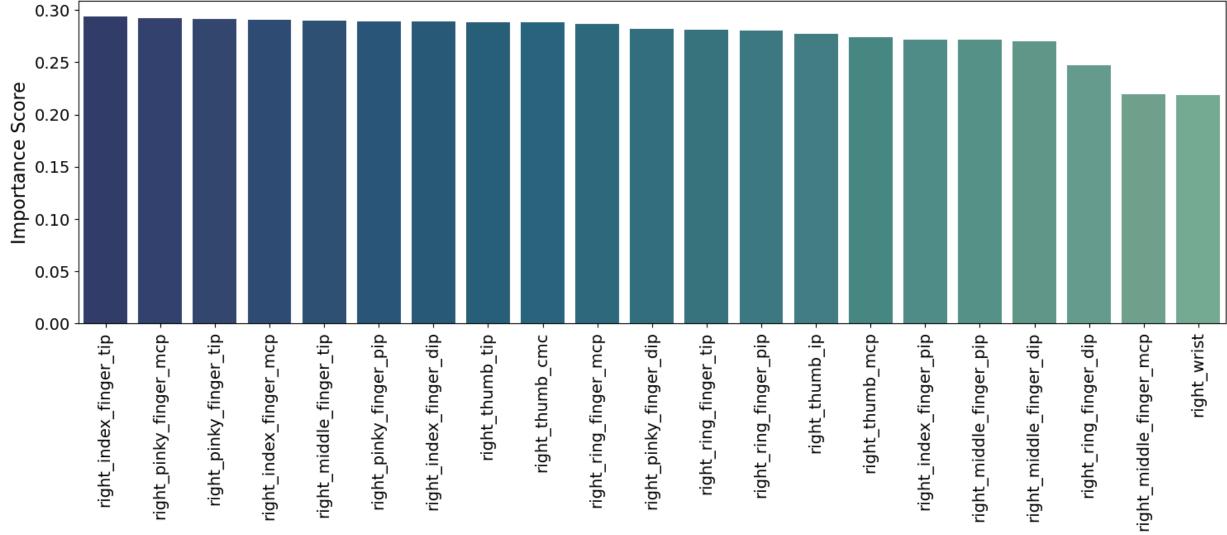


09

#### **Class Distribution**

## **RIGHT HAND KEYPOINT IMPORTANCE**

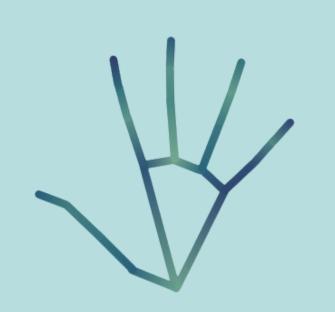


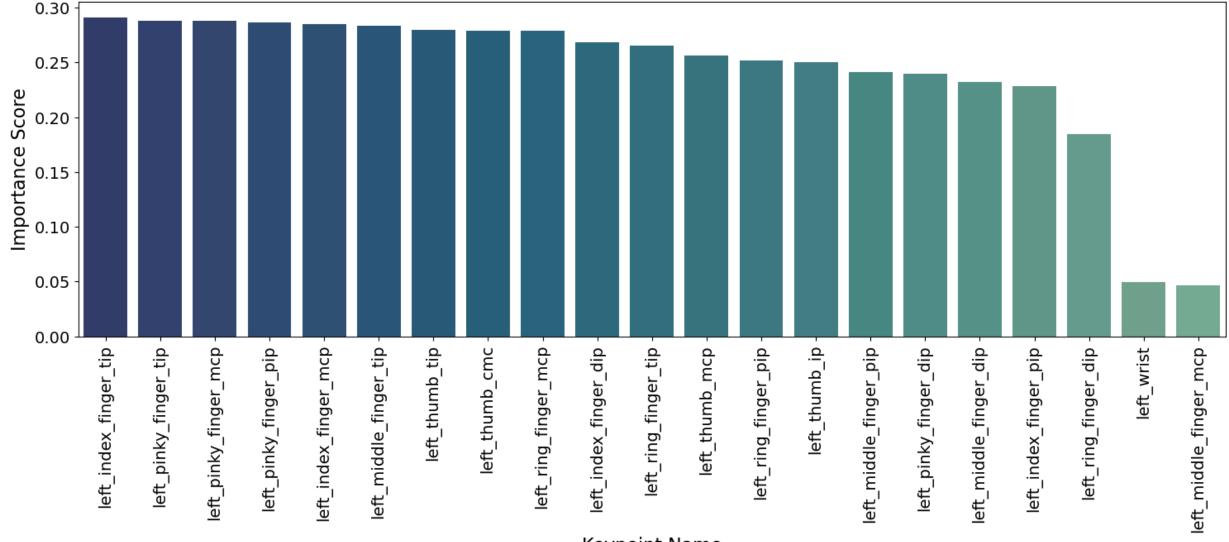


10

Keypoint Name

## **LEFT HAND KEYPOINT IMPORTANCE**





 $\bullet \bullet \bullet$ 

11

Keypoint Name

## HAND KEYPOINT IMPORTANCE

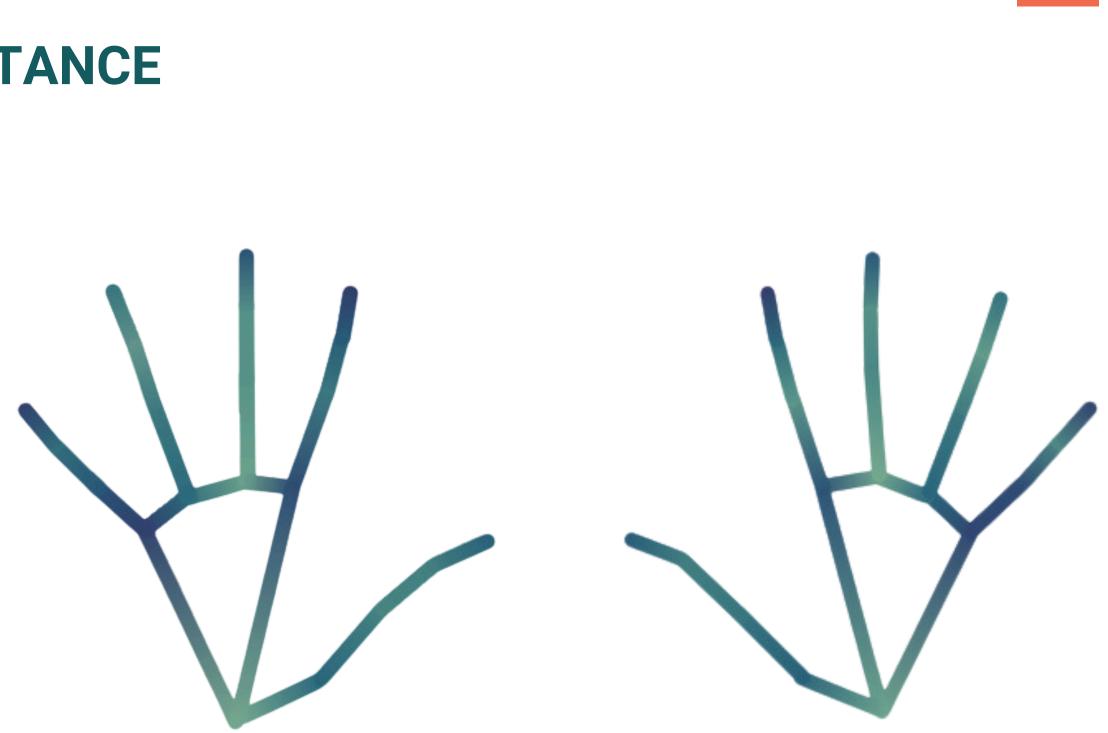
#### **INNER FINGERS**

Higher level of occlusion due to their position.

**INNER JOINTS** Higher level of occlusion or lack of accurate depth information.

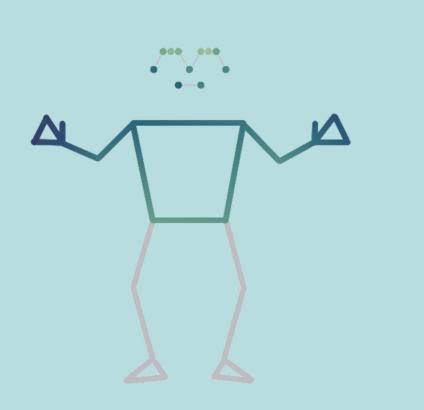
#### **VARYING IMPORTANCE**

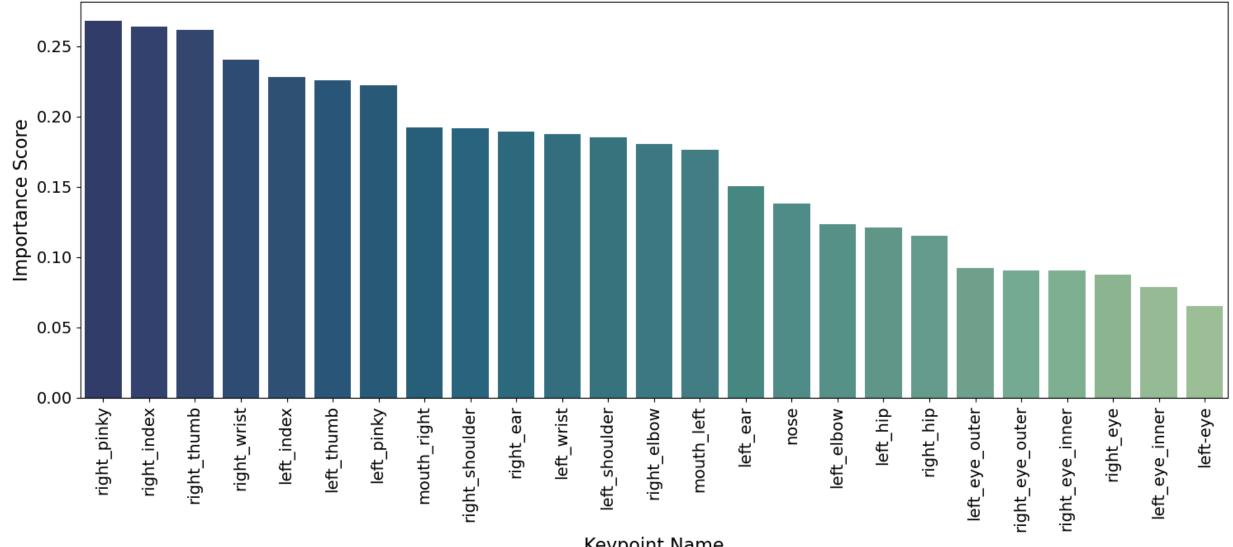
Potential bias if not featured predominantly in a large number of signs.





### **BODY KEYPOINT IMPORTANCE**

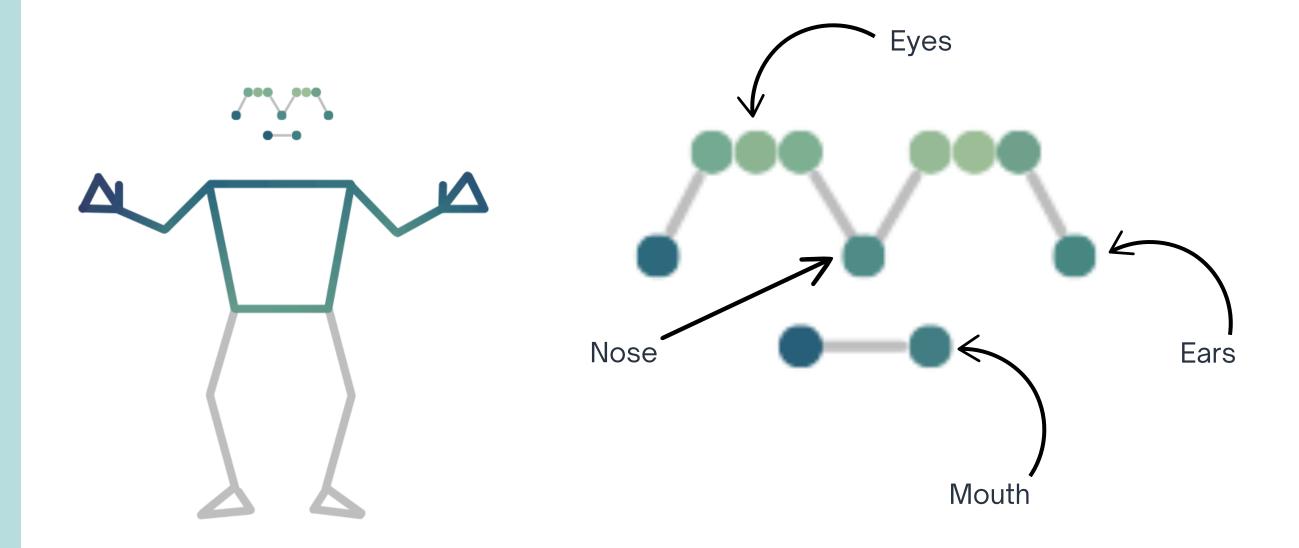




13

Keypoint Name

### **BODY KEYPOINT IMPORTANCE**





# **KEY OBSERVATIONS**

**Keypoint distortion** - occlusion and lack of accurate depth information.

Gloss imbalance - over-reliance on keypoints prominent in large number of signs.

Facial features - surprisingly under-utilised.

Feature importance for feature selection - feature importance does not always equate to linguistic importance.

# **LIMITATIONS & FUTURE WORK**

Single language dataset - expand to include additional datasets for more language-agnostic evaluation.

Single SLR model architecture - additional models to reveal overlapping trends in feature utilisation.

17

# **THANK YOU**

