

#### **ORLian:**

A novel method to help predict survival from HPV-induced oral cancer, by training mutli-modal convolutional network based on siamese similarity pretraining model with anonymised patient pathology data and other histological data

## **Methods**

Histological sections from biopsies were used to train the El-nAl-ny model, based on a pre-trained Res-Net\*, then visualized using GRAD-CAM \*\*

Training was done using data stored on Dataiku and generated images were compared back against image stack analyses

#### Introduction

While people still are worrying about waves of Covid-19 cases, tsunamis of cancer cases globally are worrying.





2020 est Total incidence Total mortality Incidence and mortality maps from https://gco.iarc.fr/

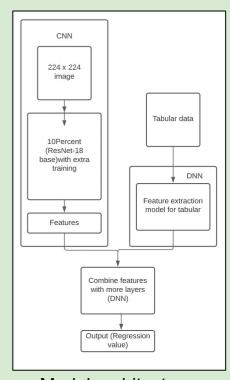
Ear Nose and Throat (ENT) cancers caused by Human Papilloma virus (HPV) spread as a sexually transmitted disease, and we need better tools to prevent mortality in patients & database access for open science.

Artificial intelligence may help.

### **Evaluation Methods**

Model evaluation was done using two methods:

- Using XGBoost and SHAP to determine feature importance from the tabular
- 2) Use Leave One Out to remove one marker feature at a time and checking performance
  - Medical expertise evaluate the model and rank their layers too



Model architecture

**Future Prospects** 

Model optimization and explainability

More clinical data

Model deployment

Clinical trials also in

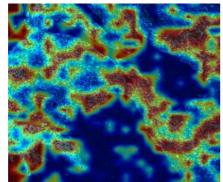
light of possibilities for

### Results

**Current Score** on Macro: 5.766

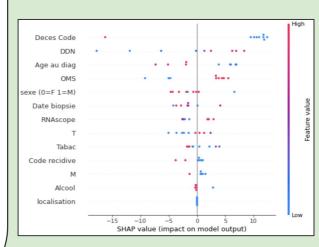
Model outperforms many publicly available models such as ResNet-18

Can capture cellular structure in images better than original baselines



**GRAD-CAM** visualisation

Leave One Out: Average performance on validation sets across folds for markers



Feature importance from the tabular



# **Acknowledgements**

vaccination now

JOGL Epidemium 3 ORL/IA challenge @Nicolas Hegerle of JEDHA

# ORLIAN CORE TEAM Vikramaditya Chhapwale

Omar Osman Rachel Aronoff

#### **Feature and Layer Importance:**

Through Leave-One-Out and manual evaluation using medical expertise, we find that in terms of marker importance:

#### 5 & 3 > 1 & 2 > 6 & 4

Feature importance from the tabular can be ranked

- 1. Whether the person has died
- 2. How old they were
- Age of diagnosis 3.
- Severity of condition 4.
- 5. Gender:
- \* Ciga, O., Xu, T. and Martel, A.L., 2022. Self supervised contrastive learning for digital histopathology. Machine Learning with Applications, 7, p.100198.
- \*\* Selvaraju, R.R., Cogswell, M., Das, A., Vedantam, R., Parikh, D. and Batra, D., 2017. Grad-cam: Visual explanations from deep networks via gradient-based localization. In Proceedings of the IEEE international conference on computer vision (pp. 618-626).