Addressing challenges to enable better use of routinely collected clinical photographs: describing the largest cleft dataset for machine learning analysis

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Medical Artefacts



- A reliable and accurate scoring system utilising large numbers of unstandardised 2-dimensional photographs of ethnically diverse patients, which is inexpensive, widely accepted and easily applicable, does not exist.
- Artificial Intelligence (AI) has yet to be used in cleft care.
- AI lends itself to big data and limits human bias

The results of visual assessment of a dataset of images to determine the severity of the cleft lips using an attention enabled DenseNet model. Coloured boxes identify patterns and clusters in the image.

Methods + Results

500000 Type of Orofacial Cleft The data set comprised of over 5 million photos 450000 Database collected across 90 countries in a 20 year period 400000 350000 300000 Data were described using RStudio + Data **Microsoft Excel** 250000 Description 200000 Analysis showed the dataset is 150000 Data the largest + most ethnically 100000 Analysis diverse 50000 Cleft Palate Unilateral Unilateral Bilateral Bilateral Total number Ethnicity of Key <u>Total</u> Cleft Lip + Palate Cleft lip Cleft Lip + Cleft Lip Patient of patients Numerical <u>number</u> Palate Data An isolated unilateral cleft lip was the most frequent orofacial cleft Number of 5,349,000 recorded, along with a unilateral cleft lip and palate. Patients who Asian 728,456 photos had a bilateral cleft lip without a cleft palate were less frequent Indian 625,220 Unique 1,589,080 patient IDs 122,846 **Key Points** Black Treatment 1,605,980 52.788 Latino 1. A global need for an updated and ethnically inclusive postphotographs op aesthetic scoring system exists + we need to utilize 'real White 35.552 world' data Mixed 21,416 2. The dataset is the largest global dataset comprising routinely collected photographs of patients with an orofacial Other 19,504 cleft that exists. **Pacific Island** 139 3. Some data were mislabeled, varied There were 9 types of image in the in quality and lack of minimum dataset, with 5 types of view standardization (examples to the left.) Pre-operative 4. The data are ethnically inclusive frontal images were the most 5. Numerous photos contained 'medical common artefacts' (right) complicating AI analysis Two data sets (training and Conclusions teaching) were created for The quality of data is suboptimal for research, but can be used both supervised and unsupervised ML analysis with success, yielding results that are more globally applicable and representative of the global burden of disease. Recommendations for future researc Data were chosen based on Pre-op + post-op frontal and worms eye view photography inclusion criteria (left) should be a minimum Medical artefacts should be excluded in the photographs if

Photos of patients with a cleft

palate only (left) were used as

control images and AI analysis

was run

possible

5 main criteria: Pre-op photo, frontal/worms eye image, ≤ 36 months, unilateral cleft lip only

References

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