

Facial Landmarks Predictions

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ML-based approach -

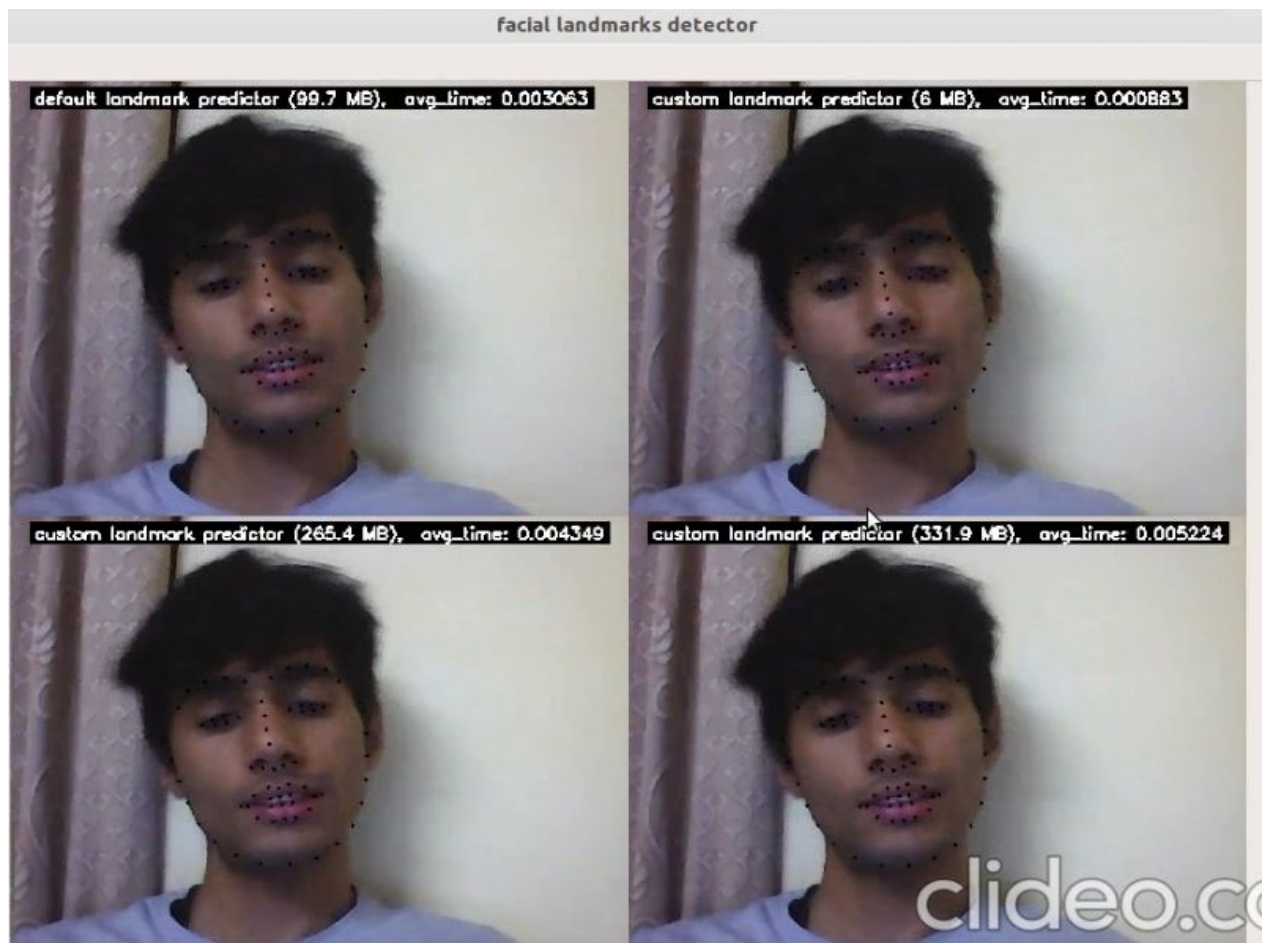
Before jumping straight to Deep learning I explored ML ways of approaching this problem and found that in fact, dlib is doing the same as well. Training a custom dlib facial landmarks predictor in dlib is much simpler than I thought. Dlib provides a very easy to use API for that. For this task, dlib utilizes the power of Ensemble models.

To estimate the landmarks location, the algorithm,

- Examines a sparse set of input pixel intensities (the features to the model)
- Passes the features into as **Ensemble of Regression Trees** (ERT)
- Refines the predicted locations to improve accuracy through a **cascade of regressors**

I trained 3 custom dlib facial landmarks predictors by changing various parameters like **cascade_depth**, **tree_depth**, **num_test_splits**, **oversampling_amount**, etc. For this, I used the **ibug_300w** dataset provided by dlib itself.

Here is a comparison of various custom and default dlib predictors,



Pros -

- Lighting fast
- Easy to train and test
- Works perfect for frontal faces.

Cons -

- Have only 68 landmarks, missing various important regions like the forehead which are important for many tasks (we can fix this by annotating new points and using more data).
- It doesn't perform well for left and right profile faces, also for faces with larger tilts (we can fix this by training more on non-frontal faces).

You can refer to [this](#) notebook for more details on the training and evaluation of the above custom models.

Refer [to this](#) video to see the model comparison in more detail.

Refer [to this](#) article for more details on dlib API for facial landmarks detection.

Deep-Learning based approach -

The task of predicting landmarks is very well achieved by deep-learning based approaches and outperforms traditional approaches in terms of performance. The simplest of them is to use any well known CNN model like **VGG**, **ResNet**, **DenseNet**, **InceptionNet**, etc for regressing the landmarks.

Apart from this, I found [this](#) MTCNN (Multi-task Cascaded CNN) which not only finds faces in an image and also 5 key landmarks on the face (2-2 for eyes and lips and 1 for nose tip). For this it uses a cascade of 3 CNNs, they call them 3 stages names as **P-Net**, **R-Net**, **O-Net**. It gives really good results in real-time.

Refer [to this](#) article for more details on MTCNN.

Pros -

- Takes more time as compared to dlib
- Deep learning approaches suits this kind of complex tasks

Cons -

- Currently supports only 5 facial landmarks (but we can extend this by annotating and using more data)