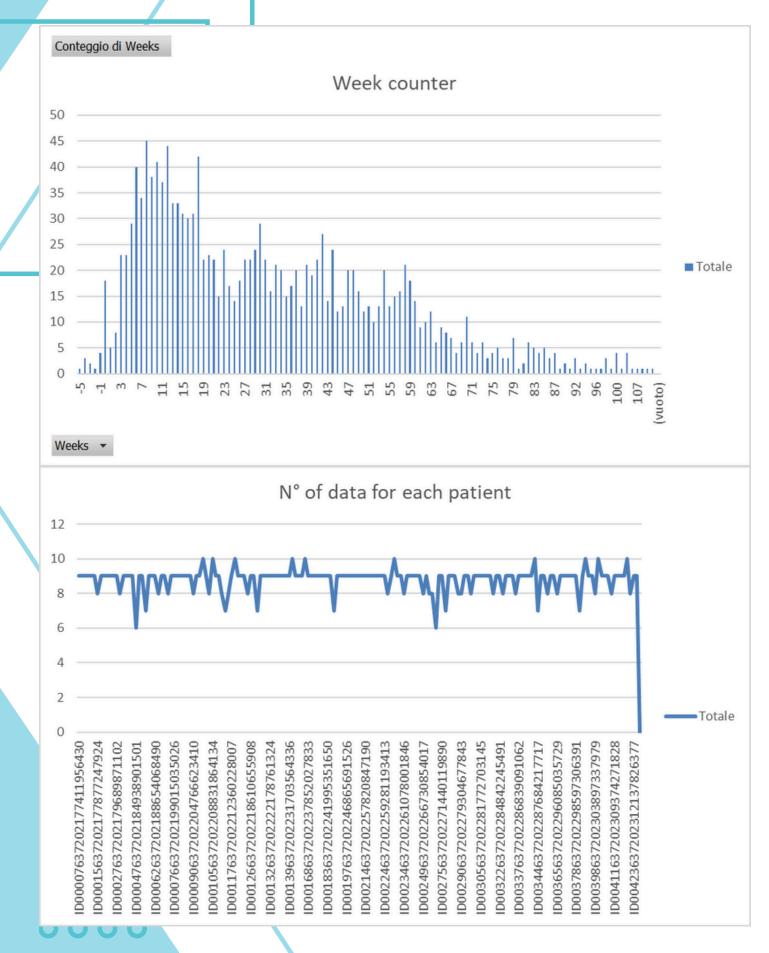


# PROGRESS PREDICTION 3°WEEK





# Data analysis



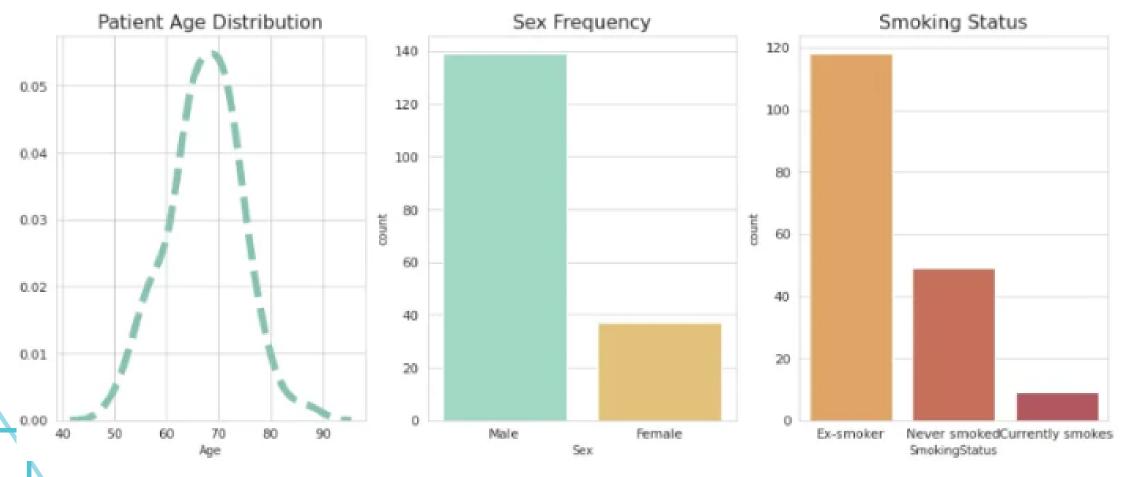
#### Week analysis:

• 18 patients have data at time 0

0000

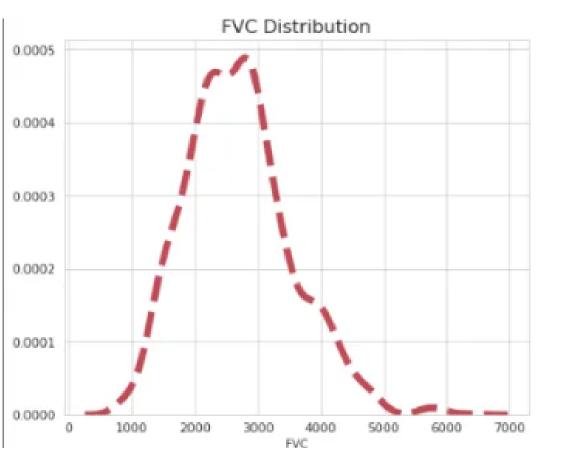
- Week 8 is the most present with 45 pairs
- Most patients have more or less same amount of data (176 distinct patients)

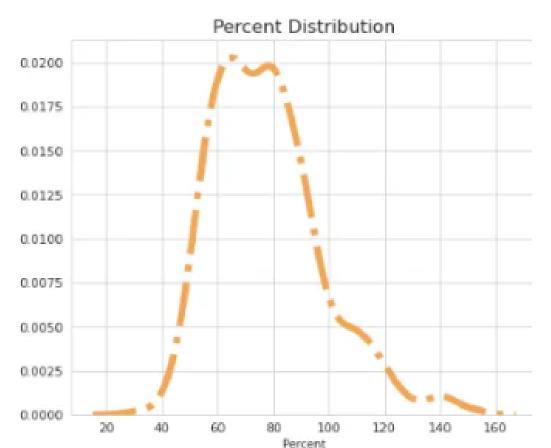
# Data analysis



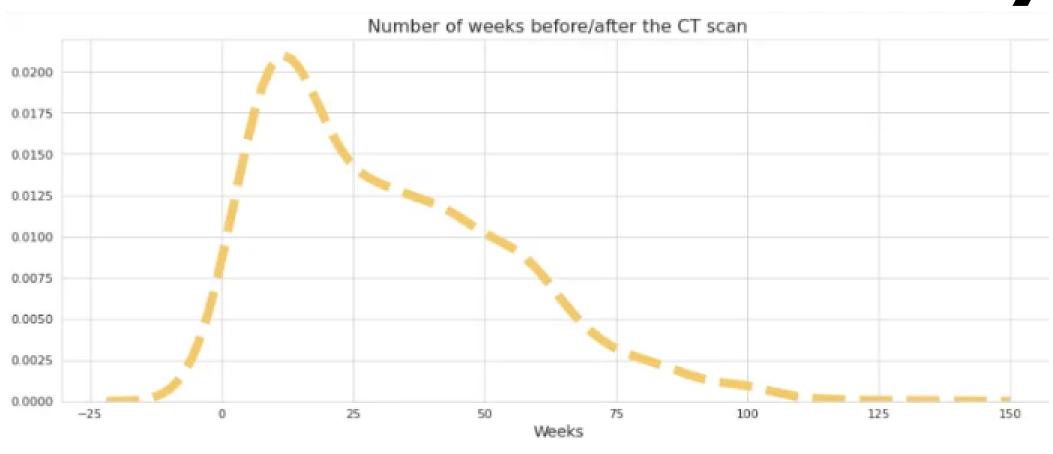
• Patient age mean → 67 years

- Male most prevalent
- Ex-smoker most prevalent





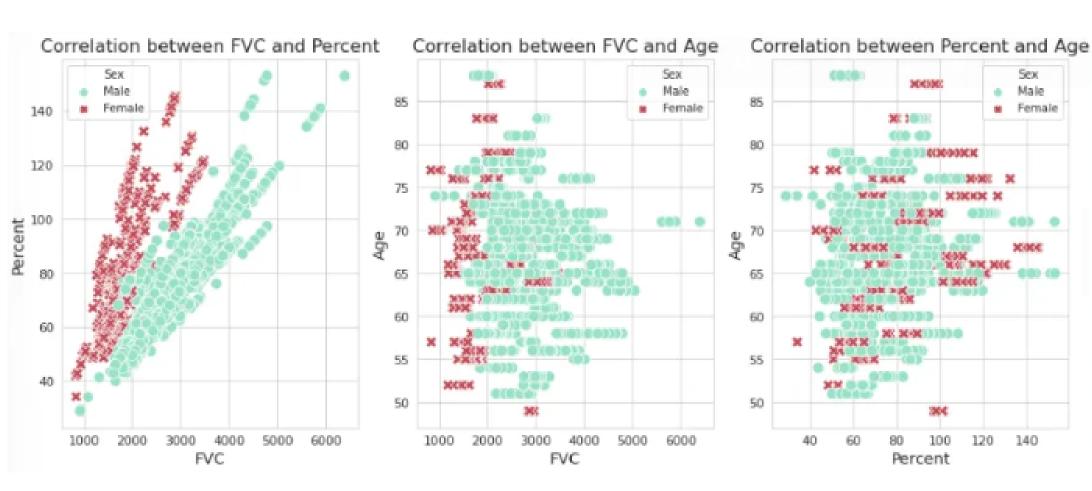
# Data analysis

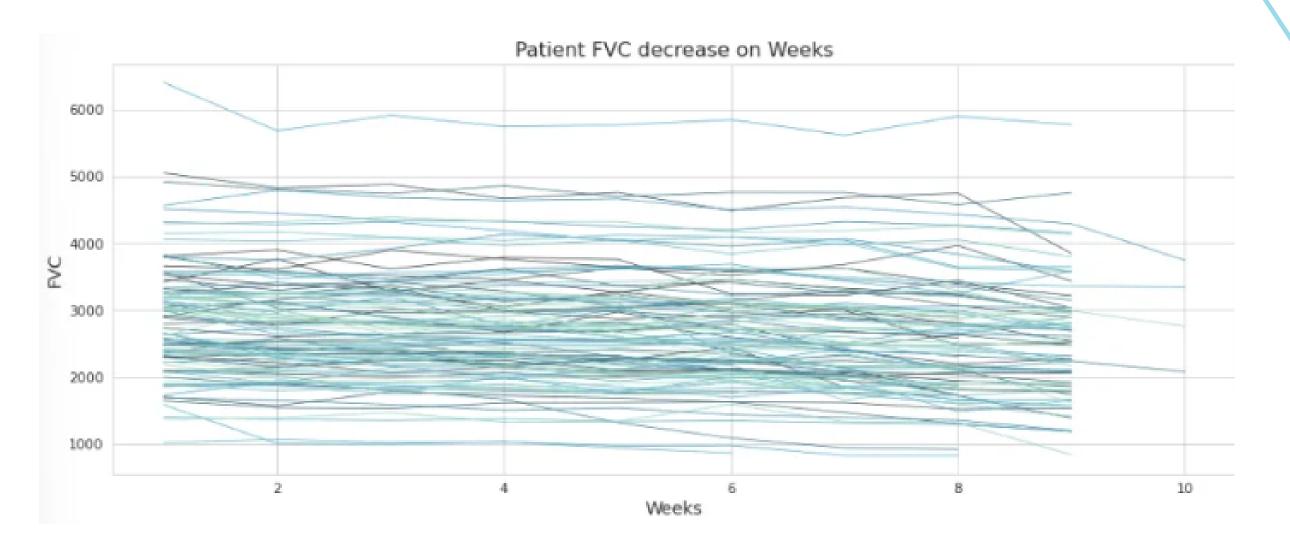


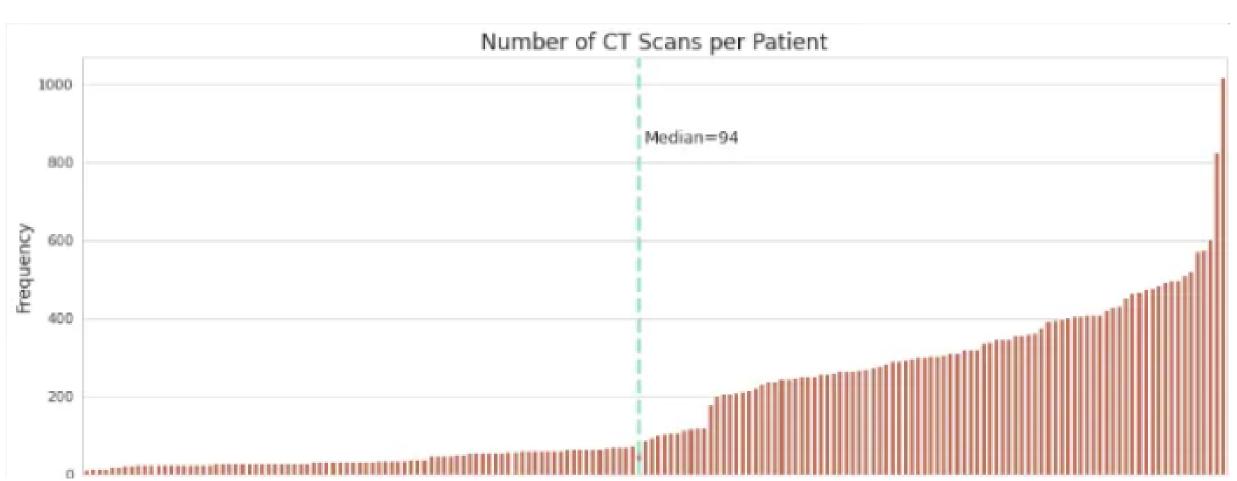


- PEARSON CORR FVC X PERCENT: 0.672
- PEARSON CORR FVC X AGE: -0.09
- PEARSON CORR PERCENT X AGE: 0.096









## **KAGGLE SOLUTION**

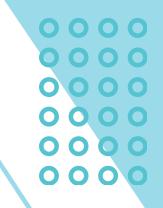
0000

11 images from the middle 30-60 percentile for each CT scan. Using this the following features were extracted:

- Lung volume
- Average of number of tissue pixels present across all selected images for a patient
- Average tissue area
- Average tissue volume
- Average of (number of tissue pixels)/(Total pixels on image)
- Average of (tissue area)/(lung area)

Then merge image and metadata, clean outliers and normalize data.

## KAGGLE BEST SOLUTION



## Model ensemble



+ Quantile Regression

### KAGGLE BEST SOLUTION



#### 1<sup>st</sup> model

- Linear Regression → Take as input (Weeks,FVC) and determine a slope "a"
- Slope "a"  $\rightarrow$  Indicates the progression through the weeks

The slope is then used as a target for the pair (CT scans, data)

CNN (EfficientNet) → Take as input CT scan and data

#### 1st model



K-fold cross-validation: 5 folds

Quantiles pre-defined for each fold as [0.8, 0.5, 0.1, 0.1, 0.1]

#### For each fold:

The model makes predictions for each patient, 20 CT slices are selected, for each slice a slope is determined and the quantile is used to aggregate the per-slice predictions.

The FVC at each week is predicted using the formula:

$$FVC = a^*w + B$$

(B is the intercept calculated based on the known FVC at week 0.

#### 2<sup>nd</sup> model

0000

• MLP

Takes as input only the available data + metadata.

Doesn't take in consideration features extracted from the CT scan.

#### Model ensemble

The two predictions are combined giving more importance to the 2<sup>nd</sup> model (0.6).

- 1st model: Gains complex information and patterns from the CT scans
- 2<sup>nd</sup> model: Stabilize the predictions

#### **Final Output:**

PatientID - Week - FVC - Confidence