Top 10 - on DATA

- 1. Ensure LOTS OF useful quality data, either by having it or getting it.
 - 2. Plot and understand input data and its most basic relations.
- 3. All data is flawed. Make sure you know how and correct/adapt/exclude.
 - 4. Map out any missing data and decide wether to filter or impute.
 - 5. Split data (Train, Valid, and Test), to not (unknowingly) overtrain.
 - 6. Consider number of features, and omit those which don't contribute.
- 7. Fast computing and access to data for quick analysis turn-around is key.
 - 8. **If data is in shortage,** consider methods for augmenting it.
 - 9. **Test several methods**, as different methods apply to different cases.
 - 10. **Make cross checks**. All data analysis results can be bugged/flawed.

Top 10 - on ALGORTIHMS

- 1. **Use appropriate model/architecture** that matches the data/problem.
- 2. Think carefully about the loss function, i.e. what you want to optimise.
 - 3. Tree algorithms are good for getting fast results on structured data.
 - 4. Neural networks are more versatile, but harder to train.
 - 5. **Variable transformation** is required for Neural Networks.
 - 6. **Image analysis** is mostly done with a Convolutional Neural Network.
 - 7. **Dimensionality reduction** benefits very high dimensional problems.
 - 8. Geometric data can be analysed with Graph NNs and Transformers.
 - 9. **Unsupervised learning/clustering** results can be difficult to interpret.
 - 10. **Uncertainties in regression** can be given by ML algorithms.