Top 10 - on DATA

- 1. Check and understand input data and its most basic relations.
- 2. Split data (Train, Valid, and Test), to not (unknowingly) overtrain.
- 3. Consider number of features, and omit those which don't contribute.
- 4. Ensure LOTS OF useful quality data, either by having it or getting it.
- 5. Fast computing and access to data for quick analysis turn-around is key.
 - 6. **If data is in shortage,** consider methods for augmenting.
 - 7. **All data is flawed**. Make sure you know how and filter it first.
 - 8. Map out any missing data and decide wether to filter or impute.
 - 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 performant and versatile, but harder to train.
 - 5. Variable transformation is typically required for Neural Networks.
 - 6. **Image analysis** is mostly done with a Convolutional Neural Network.
 - 7. **Dimensionality reduction** benefits very high dimensional problems.
 - 8. Streams of data (e.g. text) can be analysed with LSTM/RNN networks.
 - 9. Unsupervised learning/clustering results can be difficult to interpret.
- 10. **Uncertainties in regression** can be given by ML (typically NN) algorithms.