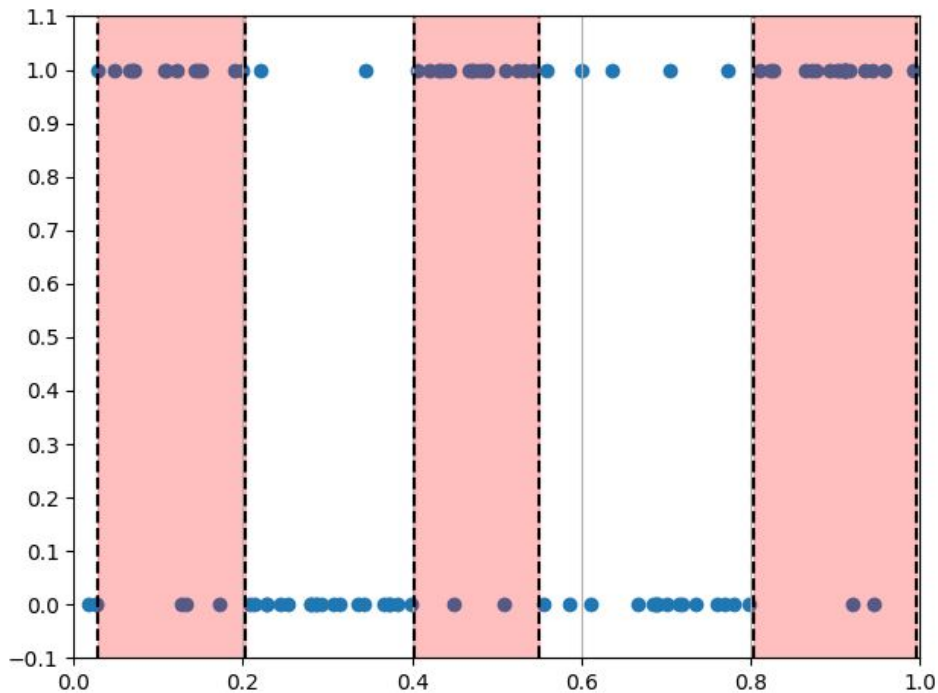


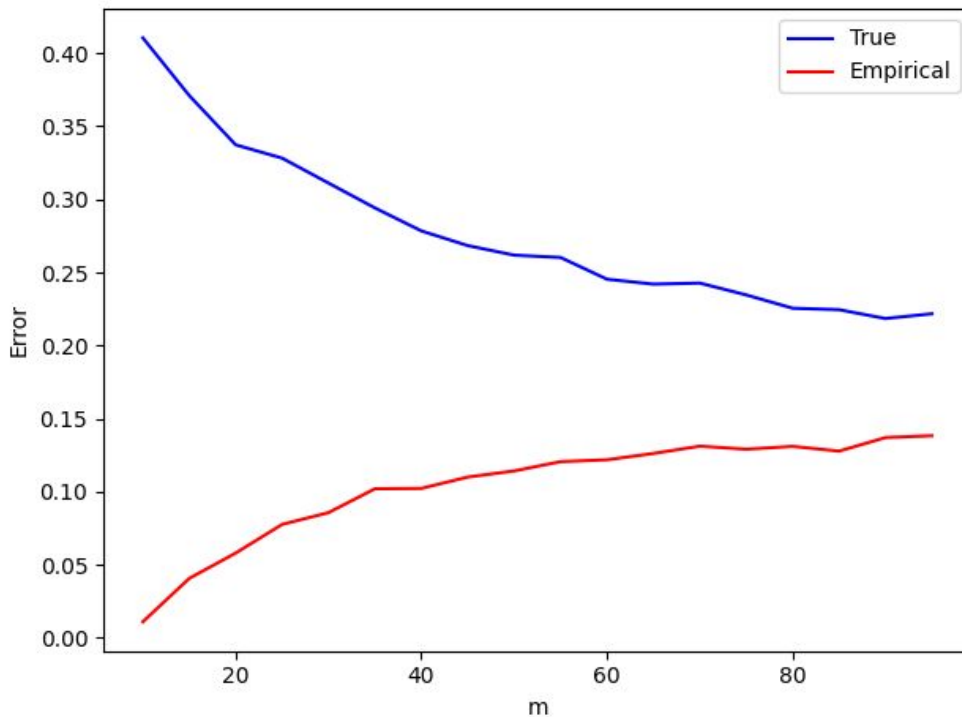
1) The red vertical stripes are the intervals provided by the find_best_interval:



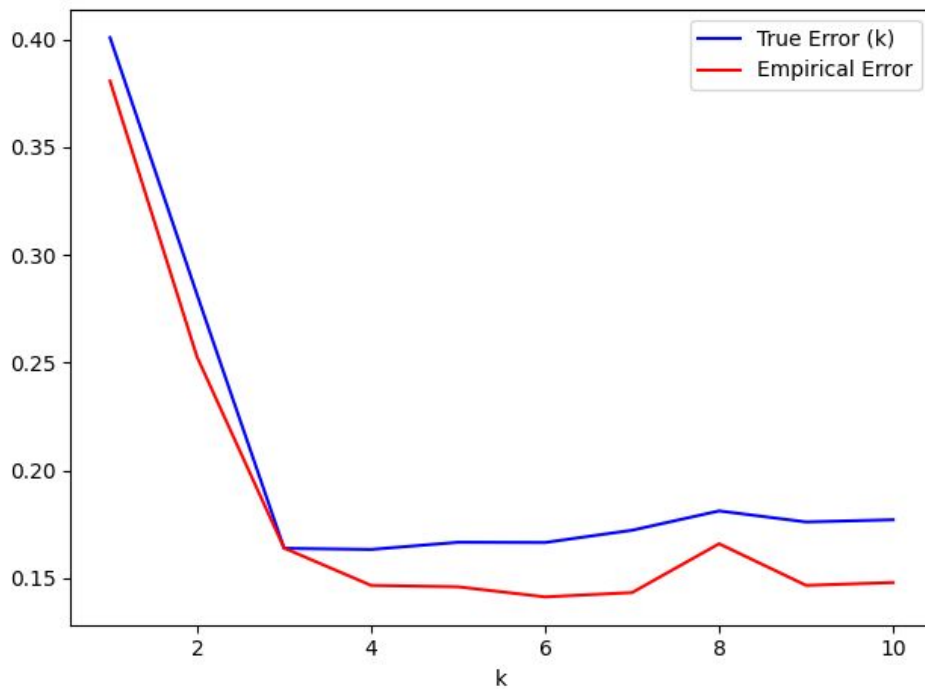
2) The best hypothesis is: $h(x) = 1$ if x is in one of the intervals $\{(0, 0.2), (0.4, 0.6), (0.8, 1)\}$, 0 else

3) As seen in the following chart the true error decreases as m increases which makes sense for two reasons:

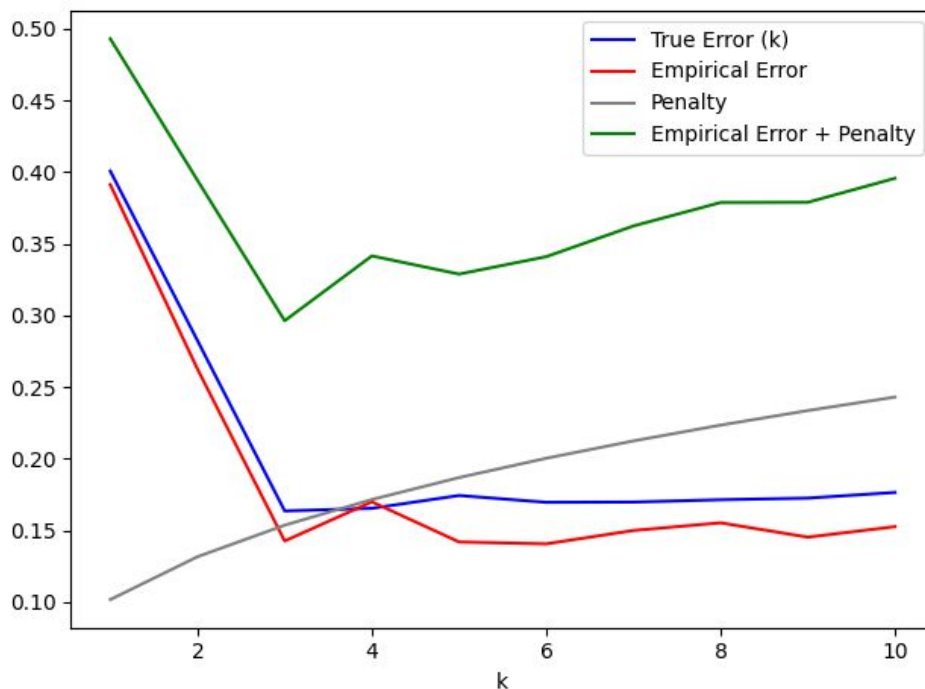
- The true error should converge to its minimum as m goes to infinity, while the empirical error should converge to the true error
- Empirical error is smaller for few examples since it's easy to train a model based on very little examples (for example if $m \approx k$)



4) k^* intervals doesn't necessarily constitute a good hypothesis, it may cause overfitting. This explains why k^* is 6 but the best k in terms of true error is 3



5) Here we can clearly see the expected result - adding the penalty function causes the best k to be 3. Without the penalty function it's still 6



6) In all 3 tests the best hypothesis was based on $k=3$. So to find the optimal hypothesis itself all we needed is to run `find_best_intervals` with $k=3$ and a big enough sample