LeetCode difficulty estimator

Neural networks semestral project

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LeetCode

- popular platform for developers
- rich community
- content
 - Contests
 - Mock interviews for jobs
 - Problems/Tasks
 - Forum discussions
 - Courses

Our task

- Given a description of a problem, predict its difficulty
 - Easy / Medium / Hard

1. Two Sum

Easy ₺ 54151 ♀ 1803 ♡ Add to List ₺ Share

Difficulty (target)

Given an array of integers nums and an integer target, return indices of the two numbers such that they add up to target.

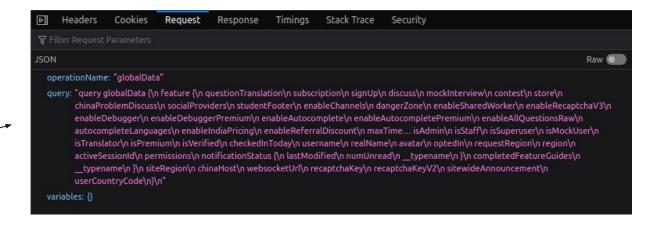
You may assume that each input would have exactly one solution, and you may not use the same element twice.

You can return the answer in any order.

Textual description (data)

How to obtain the data?

Query the LeetCode internal API





Querying the API

```
graphql_request_problems_list = leetcode.GraphqlQuery(
   query="""
        query problemsetQuestionList($categorySlug: String, $limit: Int, $skip: Int, $filters: QuestionListFilterInput) {
           problemsetQuestionList: questionList(categorySlug: $categorySlug, limit: $limit, skip: $skip, filters: $filters) {
               total: totalNum
               questions: data {
                   acRate
                   difficulty
                   title
                   titleSlug
                    topicTags {
                       name
                       id
                       slug
                    hasSolution
    variables=variables.
    operation_name="problemsetQuestionList"
```

1. Query the list of problems names

2. For each problem query the data

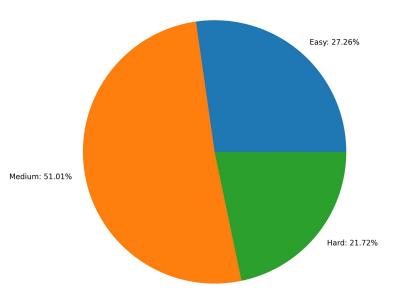
For each problem, query the difficulty and description.

result = {}

Dataset

- 2950 problems
 - ~584 problems for premium users
- 2366 examples to train

Distribution of Problem Difficulties



Understanding training examples

"content": "Given an array of integers <code>nums</code> and an integer <code>target</code>, return indices of the two numbers such that they add up to <code>target</code>.\n\nYou may assume that each input would have exactly one solution, and you may not use the same element twice. \n\nYou can return the answer in any order.\n\n \n<strong class=\"example\">Example 1:\n\n\nInput: nums = [2,7,11,15], target = 9\n0utput: $[0,1]\nExplanation: Because nums[0] + nums[1] == 9, we return [0, 1].\n\n\n<strong$ class=\"example\">Example 2:\n\n\nInput: nums = [3,2,4], target = 6\nOutput: [1,2]\n\n\n<strong class=\"example\">Example 3:\n\n\nInput: nums \n\n\t<code>2 <= nums.length <= 10⁴</code>\n\t<code>-10⁹ <= nums[i] <:= 10⁹</code>\n\t<code>-10⁹ <:= target <:= 10⁹</code>\n\t> Only one valid answer exists.
</rr>
/ul>\n\n\n\n\strong>Follow-up: Can you come up with an algorithm that is less than (<code>O(n²)</code>font face=\"monospace\"> time complexity?" Sometimes there are examples mentioned Sometimes required complexity is mentioned

Lots of HTML clutter

- What is relevant? What should be considered as features?

Primary text feature extraction

- Tf-ldf
 - Word-level & character-level ngrams
- Word embeddings
 - Static
 - GloVe
 - fastText
 - Contextualized
 - BERT

$$w_{x,y} = tf_{x,y} \times log(\frac{N}{df_x})$$



 $tf_{x,y}$ = frequency of x in y df_x = number of documents containing x N = total number of documents

Shallow learning models

- 80:20 train:test
- Tf-idf
 - ~400k features (1-3 word-level, 1-5 character level)
- class imbalance problem
 - Oversampling (SMOTE)
 - balanced batches
 - downsampling

Classification

(1 out of 3 classes)

- Perceptron
- SVM linear kernel
- MLP Classifier

Regression

(prediction on a scale from 0 to 2)

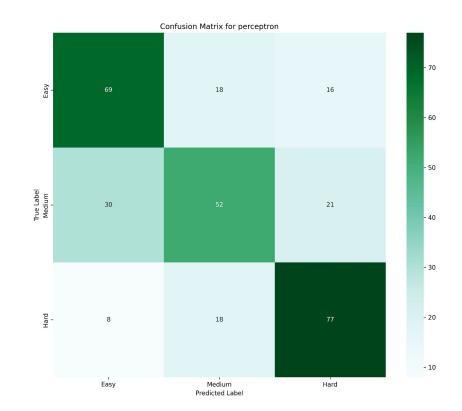
MLP Regressor

Perceptron

Best model:

- Tf-idf
 - 1 to 5 character level n-grams
 - 1 to 3 word level n-grams
- SelectPercentile(percentile=50)
- *L2* penalty
- alpha 0.0001

Stratified k-fold, k = 5 Average test macro f1: **58.26** Average test micro f1: **58.56**

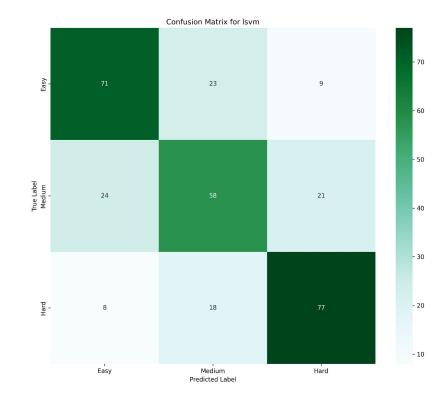


Linear SVM

Best model:

- Tf-idf
 - 1 to 5 character level n-grams
 - 1 to 3 word level n-grams
- SelectPercentile(percentile=5)
- Square hinge loss
- L2 penalty
- alpha 0.0001

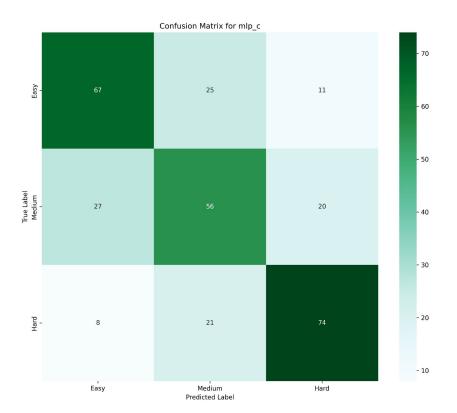
Stratified k-fold, k = 5 Average test macro f1: 60.94 Average test micro f1: 61.15



MLP Classifier

- Adam optimizer, momentum 0.9 with Nesterov, ReLU, Tf-Idf word & character 3-grams, 5-grams
- Stratified k-fold, k = 5
- SelectPercentile(percentile=5)

Hidden	Average	Average
layer(s)	test	test micro
size	macro f1	f1
<u>(64.)</u>	<u>59.42</u>	<u>59.61</u>



MLP regressor

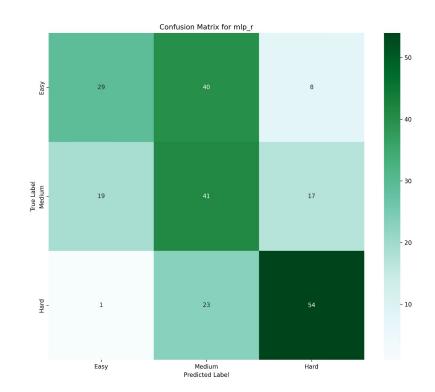
linear scale

0 - easy, 1 - medium, 2-hard

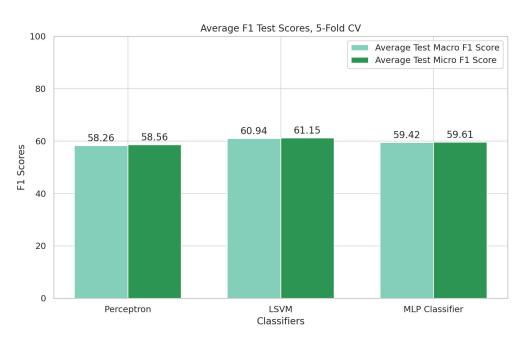
rounding the predictions 0.75, 1.25 thresholds

Test RMSE: 0.675

Test accuracy: **53.4%**



Shallow learning summary



Inference problems

- Difficulty may seem subjective
 - Easy vs. Medium / Medium vs. Hard?
- We want to minimize Easy vs. Hard problems
 - Easy problems predicted being Hard
 - Hard problems predicted being Easy
- An example of a common problem
 - Hard to classify "min-max" problems correctly

2078. Two Furthest Houses With Different Colors

Easy ₺ 865 ♀ 24 ♡ Add to List ₺ Share

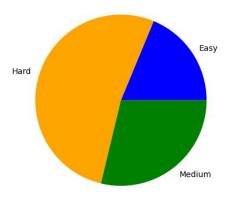
There are $\,n\,$ houses evenly lined up on the street, and each house is beautifully painted. You are given a **0-indexed** integer array colors of length $\,n\,$, where colors[i] represents the color of the $\,i^{th}\,$ house.

Return the maximum distance between two houses with different colors.

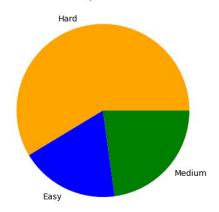
The distance between the $\,i^{th}\,$ and $\,j^{th}\,$ houses is $\,abs(i\,$ - $\,j)$, where $\,abs(x)\,$ is the $\,$ absolute value of $\,x\,$.

Predicted as hard by each classifier.

Distribution of difficulties of problems that contain "maximum"



Distribution of difficulties of problems that contain "minimum"



Classification using contextualized embeddings

- Tensorflow 2.12, KerasTuner
 - Hyperband: A Novel Bandit-Based Approach to Hyperparameter Optimization
- Pre-trained bert-base-uncased
 - No fine-tuning
- General training information
 - Batch size 32
 - Datasets sizes 1079:231:232 samples (70:15:15)
- On average, contextualized representations are more context-specific in higher layers
 - https://ai.stanford.edu/blog/contextual/



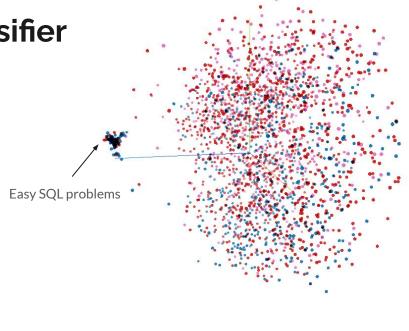






Features

- 1. Layer BERT embeddings
- mean pooling
- (768,)



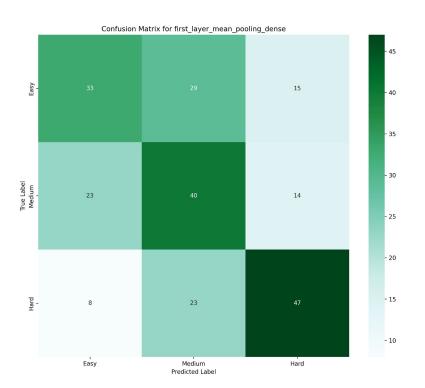






Model

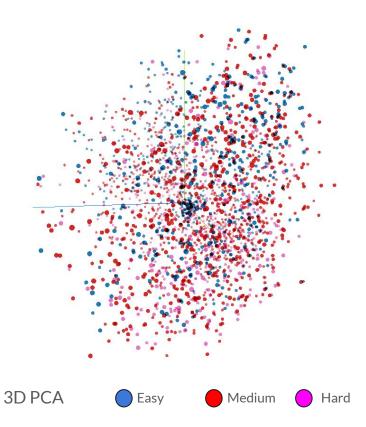
- Classification head
 - 1 x Dense
 - 576 (relu)
 - Adam optimizer
 - no Batch Norm
 - initial Ir 0.01 with cosine decay
 - dropout 0.4
 - L1~0.00004
 - L2 ~0.0003



Test accuracy: **51.7%**

Features

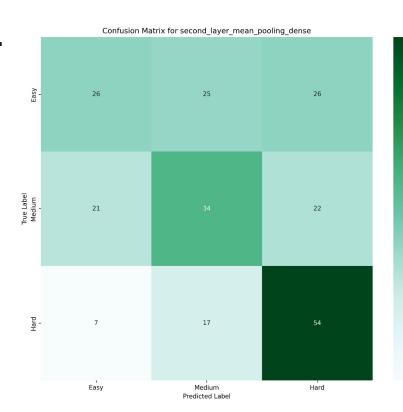
- 2. Layer BERT embeddings
- mean pooling
- (768,)



Model

- Classification head
 - 1 x Dense
 - 960 (selu)
 - SGD optimizer, momentum 0.9 with Nesterov
 - no Batch Norm
 - initial Ir 0.01 with cosine decay
 - dropout 0.1
 - L1 ~2.25
 - L2 ~1.27

Test accuracy: 49.1%



Improvements

- Custom tokenization
 - E.g. handle mathematical expressions
- Fine-tune LLM first, then extract embeddings
- Try other models like ELMo, GPT-2 for comparison
- In-context learning classification

Showcase



You are given an integer n. Return the sum of first n fibonacci numbers.

Constraints:

1 <= n <= 100

Best model prediction: Easy



You are given a hard task. Compute average of 2 numbers a and b.

Best model prediction: **Easy**



I am losing my interest in human beings; in the significance of their lives and their actions. Someone has said it is better to study one man than ten books. I want neither books nor men; they make me suffer. Can one of them talk to me like the night – the Summer night? Like the stars or the caressing wind? The night came slowly, softly, as I lay out there under the maple tree. It came creeping, creeping stealthily out of the valley, thinking I did not notice. And the outlines of trees and foliage nearby blended in one black mass and the night came stealing out from them, too, and from the east and west, until the only light was in the sky, filtering through the maple leaves and a star looking down through every cranny.

You are given an integer n. Return the sum of first n fibonacci numbers. Constraints: 1 <= n <= 100

Best model prediction: **Medium**

Thank you for your attention!