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**CalcGPT Report**

**I. Method for encoding strings (1 point)**

Baseline Strategy Input string = "x1 + x2 = "

New Strategy Input string = "2 + 3 = 5, 7 + 8 = 15, 9 + 10 = 19, x1 + x2 =\n"

Following are the examples of the outputs encode input function for baseline and new strategy. For these examples, I have created a dataset with start\_id = 0 and end\_id = 1

The dataset is small so I can completely display the outputs here.

```
create_dataset(0, 1)
```

Outputs of the encoding string function and tokenization of these strings :

### A. Example of the Baseline input string

```
Encode strings function outputs:
input strings for the baseline strategy:
['0 + 0 = ']
```

### B. Example of the new strategy input string

```
input strings for the new strategy:

['2 + 3 = 5, 7 + 8 = 15, 9 + 10 = 19,  \n 0 + 0 = ']
```

### C. Example of the Tokenization of the baseline strategy input string

```
Input strings are:
['0 + 0 = ']
input strings after tokeniztion :
{'input_ids': tensor([[ 15, 1343,  657,  796, 220]]), 'attention_mask': tensor([[1, 1, 1, 1, 1]])}
```

#### D. Example of the Tokenization of the new strategy input string

[illegible]

In the tokenization output (C, D):

- **Input\_ids:** are the indices corresponding to each token in the sentence.
- **Attention\_mask:** indicates whether a token should be attended to or not.

Tokenization of the input strings is done using **GPT2Tokenizer**.

## II. Method for generating text (1 point)

### A. Language Model Used: GPT2

1. **Size** = Large
2. **Parameters:** 1.5 billion
3. **Dataset Length:** 8 million web pages
4. **Objective:** predict the next word given all the words in the text
5. **Available in 5 different sizes:** small, medium, large, xl, and a distilled

### B. Hyperparameters chosen

1. **num\_beams: 5**, Shifts the model from greedy search to beam search
2. **Do\_sample = true**, Enables top K sampling and top P sampling but haven't done any of them
3. **Temperature = 0.8**, This value modulates the probability of the next token
4. **Pad\_token\_id = 50256**, Id of the padding token
5. **Max\_new\_tokens = 5**, Tells the model how many new tokens to generate

## III. Method for decoding strings (1 point)

example of the results of the text generate function contains also the output tokens

**baseline strategy.**

```
Generate text functions outputs for the baseline strategy:
Input strings are:
['0 + 0 = ']
input strings after tokenization :
{'input_ids': tensor([[ 15, 1343,  657,  796,  220]]), 'attention_mask': tensor([[1, 1, 1, 1, 1]])}
output of the generate function:
tensor([[ 15, 1343,  657,  796,  220,  220,  220,  220,  220,  220,  220,  220,
         220,  220,  220,  220,  220,  220,  220,  220,  220,  220,  220,
         220,  220,  220,  220,  220,  220]])
ouput strings after batch decoding:
['0 + 0 = ']
Time to generate text: 14.81301236152649
```

New Strategy

```
Generate text functions outputs for the new strategy:
Input strings are:

['2 + 3 = 5, 7 + 8 = 15, 9 + 10 = 19, \n 0 + 0 = ']
input strings after tokenization :

{'input_ids': tensor([[ 17, 1343,  513,  796,  642,   11,  767, 1343,  807,  796, 1315,   11,
                        860, 1343,  838,  796,  678,   11,  220,  220,  198,  657, 1343,  657,
                        796,  220]]), 'attention_mask': tensor([[1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
                                                            1, 1]])}
output of the generate function:

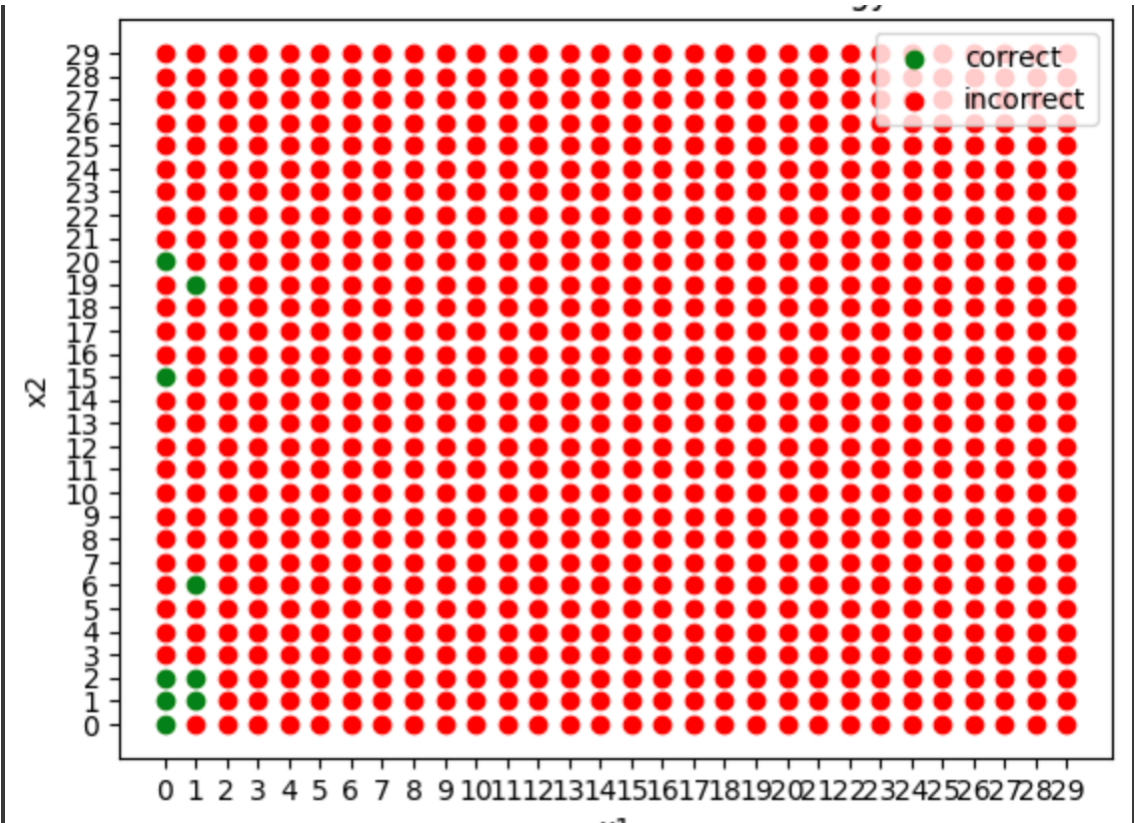
tensor([[ 17, 1343,  513,  796,  642,   11,  767, 1343,  807,  796, 1315,   11,
          860, 1343,  838,  796,  678,   11,  220,  220,  198,  657, 1343,  657,
          796,  220,  657,   11,  657, 1343]])
ouput strings after batch decoding:

['2 + 3 = 5, 7 + 8 = 15, 9 + 10 = 19, \n 0 + 0 =  0, 0 +']
Time to generate text: 2.7618303298950195
```

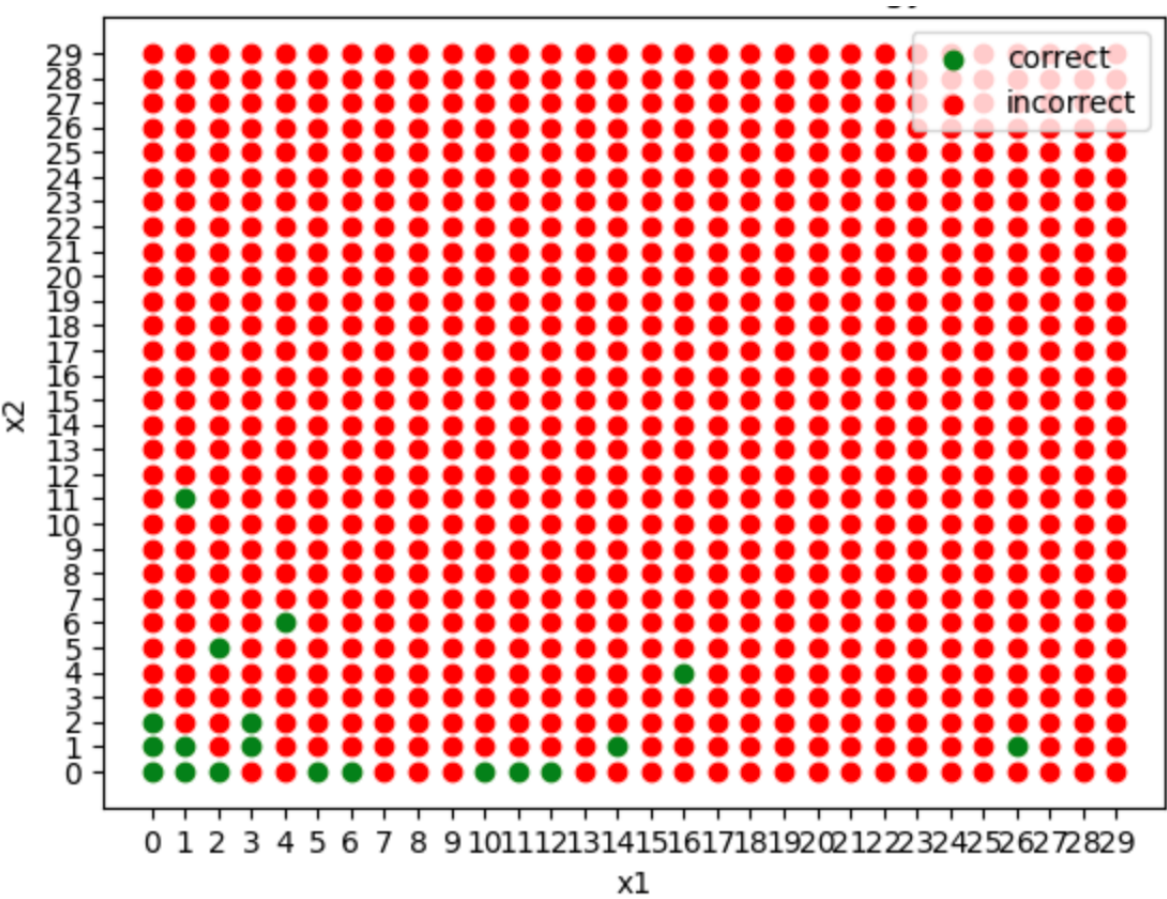
IV. Results (6 points)

**Note:** I am sharing the results of the start\_id: 0 and end\_id: 30 because I am getting good results with this as compared to start\_id: 0 and end\_id: 50 but I will be submitting i-pynb files for both start and end ids. Results will be displayed in those files aslo.

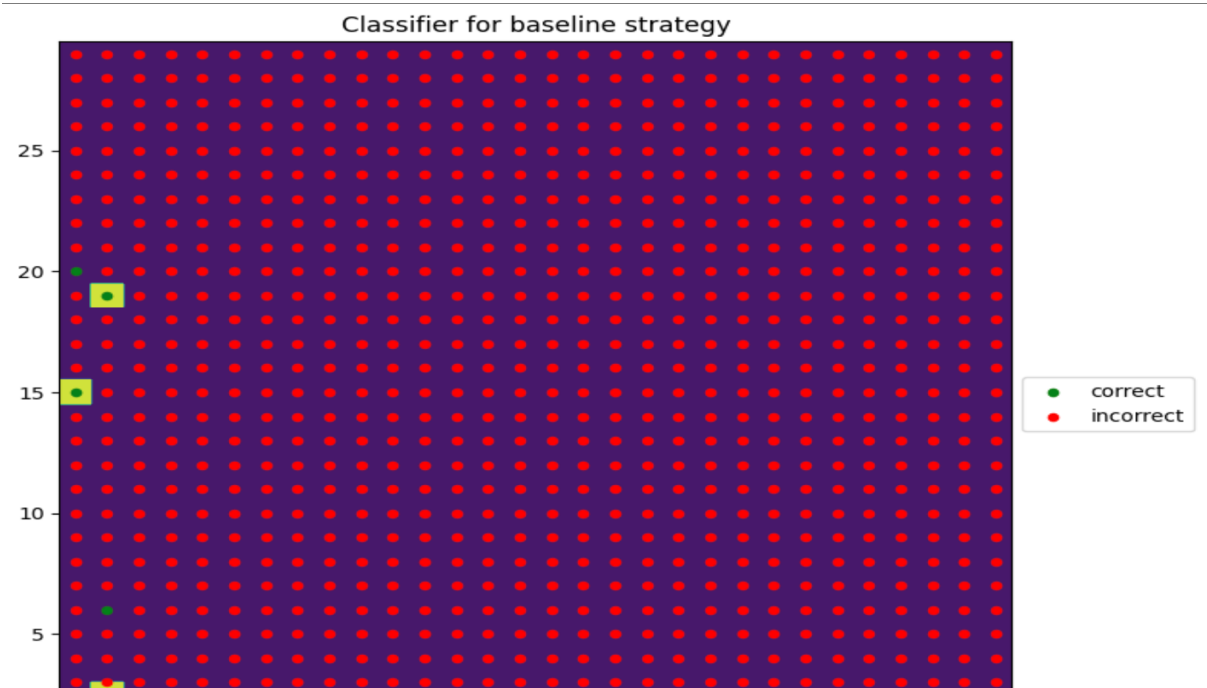
- A. LLM Accuracy for the Baseline Strategy = **007777 (0.7%)**
- B. LLM Accuracy for the New Strategy = **0.02 (2%)**
- C. Scatter Plot for the **BaseLine Strategy**.



D. Scatter Plot for the **New Strategy**



E. The Classifier Accuracy score for the baseline strategy is: 0.988 (98%)



F. The Classifier Accuracy score for the new strategy is : 0.97 (97%)

