## Exam 2 Solutions

## Question 2 - Implementing zip

```
class Zip:
    def __init__(self, l1, l2):
        self._l1 = l1
        self._l2 = l2
        self._length = min(len(l1), len(l2))
        self._curr_index = 0
    def has_next(self):
        return self._curr_index < self._length
    def next(self):
        if self.has_next():
                val = (self._l1[self.curr_index], self._l2[self._curr_index])
                self._curr_index += 1
                return val
        else:
            return None
    def reset(self):
        self._curr_index = 0
```


## Question 3 - Miscellaneous Topics

## Question 3.1 + 3.2 - Hashing

This hash function does not work.
Explanation (and a good reference answer for 3.2):
This hash function does not work because it is not consistent with the __eq function, meaning it does not return the same hash value even if they are equal objects. Two Icecream s could be equal according to _eq_ (same brand and flavor), but
could hash to different places if they had different scoops since __hash_ also uses the scoops field.

## Question 3.3

There is no single "right answer" here so we accepted any answer that answered the prompts and demonstrated a clear understanding of one of the ethical concerns we discussed in class and how it applies to the provided situation. As the criteria shows, we graded on

- Picking 1 case study to compare to
- Summarizing an ethical concern from the case study
- Comparing that case study to the provided situation
- Explanation shows depth of understanding of the problem discussed in class and how it relates to this provided situation


## Question 4 - Machine Learning

Recall that a hyper-parameter is something you specify before training the model (the choices of which impact the quality of the model that is eventually learned). The parameters of a model are the specific values learned by the model during the process of training.

## Q4.1 - Number of Hidden Layers

This is a hyper-parameter since it's something that you decide before training the model.

## Q4.2 - Number of Hidden Nodes

This is a hyper-parameter since it's something that you decide before training the model.

## Q4.3 - Weights

This is a parameter since it is learned by the learning algorithm to make the network more accurate.

## Q4.4-Bias

This is a parameter, much like the weights. This is learned by the learningalgorithm to be tuned to the specific value that works for the target task.

## Q4.5 - Activation Function

This is a hyper-parameter since it's something you specify about the network, much like the architecture (number of hidden layers / nodes). We saw a few examples of different activation functions, and which one you choose would likely lead to different models learned.

## Question 5 - Geospatial

## Q5.1-Join

Below, we show the result as a table for readability, but the specification stated we wanted your answer written as a CSV. The order of the rows/columns does not matter

## Solution

| name | continent | geometry | city | country | population | GDP |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| $\underline{W}$ | C1 | Polygon1 | A | W | 200 | 20 |
| $\underline{W}$ | C1 | Polygon1 | B | W | 100 | 50 |
| $\underline{\text { X }}$ | C2 | Polygon3 | C | X | 300 | 60 |
| $\underline{Z}$ | C1 | Polygon2 | NaN | NaN | NaN | NaN |

## Q5.2 - Plot GDP and Population

You don't need to explicitly fillna here since dissolve (like most other pandas) functions ignores missing-values in the computation (the same effect of it being a 0 for this computation)

```
fig, [[ax1, ax2], [ax3, ax4]] = plt.subplots(2, 2)
merged_country = gdf.merge(df, left_on='name', right_on='country', how='left')
```

grouped_country = merged_country.dissolve(by='name', aggfunc='sum')
grouped_continent = merged_country.dissolve(by='continent', aggfunc='sum')
grouped_country.plot(column='population', legend=True, ax=ax1)
grouped_country.plot(column='GDP', legend=True, ax=ax2)
grouped_continent.plot(column='population', legend=True, ax=ax3)
grouped_continent.plot(column='GDP', legend=True,ax=ax4)

## Question 6 - Images

Q6.1-a * b
Note: There was a typo on the exam that said $a+b$ in one place, but this doesn't have an impact on the answer since these both don't work for the same reason.

Error. Following the rules of broadcasting, b will be padded to the left with ones to become a $(1,4)$. The problem then comes from a mismatch in the second dimension where a has value 3 and $b$ has value 4 since neither of them are ${ }_{1}$ meaning neither can be stretched to match the other.

## Q6.2-Mystery 1

Either of the following shapes work

- $(5,4)$
- $(5,1)$


## Q6.3 - Mystery 3

Error. To make a 3D result, d would need to have 3 dimensions. When adding a $(a(4,3)$ ) to a 3D array, it will be

## padded on the left to a $(1,4,3)$ which cannot be broadcasted since the second and third dimensions disagree with the result shape and neither are 1.

## Question 7 - Convolution

Two common solutions are shown below

```
def color_convolution(image, kernel):
    kernel_height, kernel_width = kernel.shape
    image_height, image_width, dim = image.shape
    result_height = image_height - kernel_height + 1
    result_width = image_width - kernel_width + 1
    result = np.zeros((result_height, result_width, dim))
    for i in range(result_height):
        for j in range(result_width):
            red = image[i:i+kernel_height, j:j+kernel_width, 0]
            green = image[i:i+kernel_height, j:j+kernel_width, 1]
            blue = image[i:i+kernel_height, j:j+kernel_width, 2]
            result[i, j, 0] = np.sum(red * kernel)
            result[i, j, 1] = np.sum(green * kernel)
            result[i, j, 2] = np.sum(blue * kernel)
    return result
def color_convolution(image, kernel):
    kernel_height, kernel_width = kernel.shape
    image_height, image_width, dim = image.shape
    result_height = image_height - kernel_height + 1
    result_width = image_width - kernel_width + 1
    result = np.zeros((result_height, result_width, dim))
    for i in range(result_height):
        for j in range(result_width):
            for k in range(dim):
            curr = image[i:i+kernel_height, j:j+kernel_width, k]
            result[i, j, k] = np.sum(curr * kernel)
    return result
```

