A decorative graphic on the left side of the slide consisting of two overlapping parallelograms. The front parallelogram is blue and the back one is light green. They are both tilted at an angle, with the blue one shifted further to the left and bottom.

Neural Network Classification of Colon Cancer Histopathological Images




Presentation Outline

1. Problem
2. Data
3. Modeling
4. Recommendation
5. Next Steps

Problem



- 
- Colon cancer is the #3 most common cancer in US₁
 - 19 million colonoscopies are performed each year in the US₂
 - It can take 2-3 days to get biopsy results₃
 - Although tests aren't 100% accurate all the time, false positives and false negative results are very low. Generally is thought to occur in 1 to 2% of surgical pathology cases₄

1. Centers for Disease Control and Prevention. (2021, June 8). Colorectal cancer statistics. Centers for Disease Control and Prevention. <https://www.cdc.gov/cancer/colorectal/statistics/index.htm>

2. iData Research. (2018, August 8). An astounding 19 Million colonoscopies are performed annually in The United States. iData Research.


<https://idataresearch.com/an-astounding-19-million-colonoscopies-are-performed-annually-in-the-united-states/#::-:text=An Astounding 19 Million Colonoscopies are Performed Annually in The United States,-08%2F08%2F2018.>

3. Reasons for delays in getting your biopsy and cytology test results. American Cancer Society. (n.d.). <https://www.cancer.org/treatment/understanding-your-diagnosis/tests/testing-biopsy-and-cytology-specimens-for-cancer/how-long-does-testing-take.html>.

4. LeVea, C. (n.d.). Can a cancer biopsy result be wrong? Roswell Park Comprehensive Cancer Center. <https://www.roswellpark.org/cancertalk/202010/can-cancer-biopsy-result-be-wrong>.

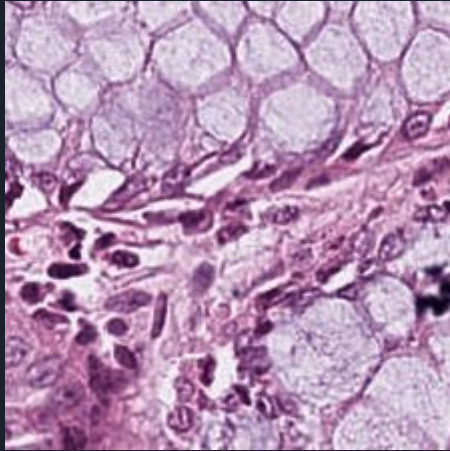
Data



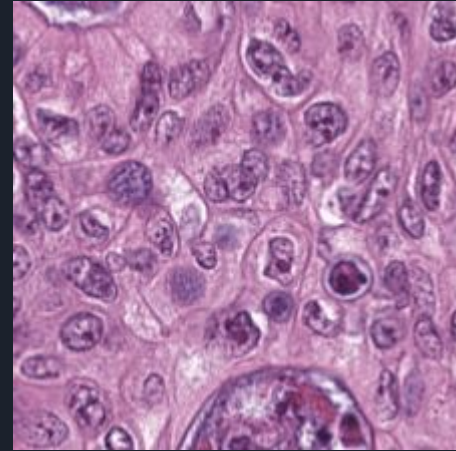
- 
- Zenodo dataset
 - Whole dataset: 100,000 images
 - 9 classes separated by a subdirectory
 - TUM (cancer cells): 14,317 images
 - NORM (normal cells): 8,763 images
 - Image size: Standard 224x224
 - Cells were preprocessed (staining)



Normal




Cancer



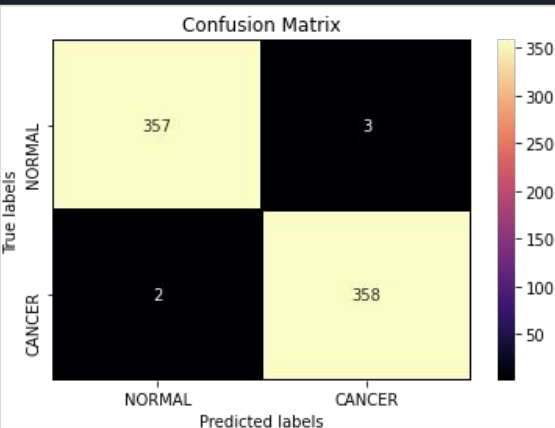
Pre-processing and Modeling



- 
- To prevent our images from losing data, the image size remained the same at 224x224
 - Before running the model, images were separated into a training, validation, and testing subdirectories.
 - The dataset was undersampled to prevent a class imbalance.
 - 8000 images for each class for training
 - 400 images for each class for validating
 - 360 images for each class for testing
 - Images were converted to JPEG from TIF
 - Training: images were augmented by adding random rotations, random brightness, random zoom range, horizontal flip, and rescaling
 - Validation: images were only rescaled
 - Testing: images were only rescaled

Modeling

- Three models were iteratively created
 - All of them severely under fitted to the images
- Transfer learning was used as the final, fourth, model
 - ResNet50: was able to predict with 99.3% precision.
 - From 720 images not used in the training process, 5 were incorrectly classified.



Recommendation

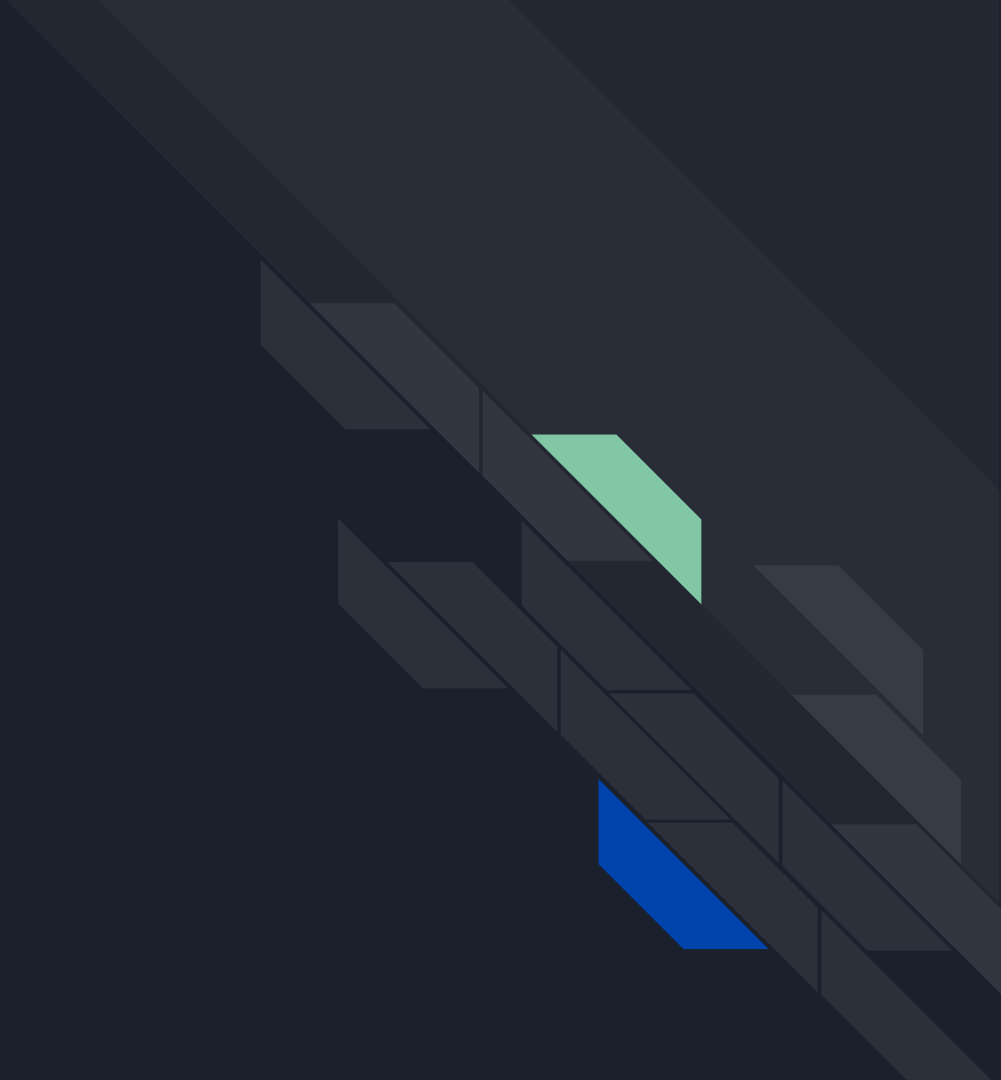




What does the model tell us

- The predictions generated by the trained model took only a little over a couple of minutes to determine 720 images.
- Due to the high efficiency and high results. This model can become an extra tool in the medical field

Next Steps





- Research the market to see where this model would be of the most help
- Use other datasets to view how the model performs
- Consider using an “unstained” version of images to see if the model holds up to the same performance



Questions?

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