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# Machine Learning HW9

## Explainable AI

ML TAs

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# Outline

- Introduction
- Topic I: CNN (hw3)
  - Model & Dataset
  - Task
  - Lime
  - Saliency Map
  - Smooth Grad
  - Filter Visualization
  - Integrated Gradient
- Topic II: BERT (hw7)
  - Task
  - Attention Visualization
  - Embedding Visualization
  - Embedding Analysis

# Introduction

# Why we need Explainable AI?

- Correct answers  $\neq$  Intelligent
- We can improve ML model based on explanation
- Make people comfortable since people need reasons

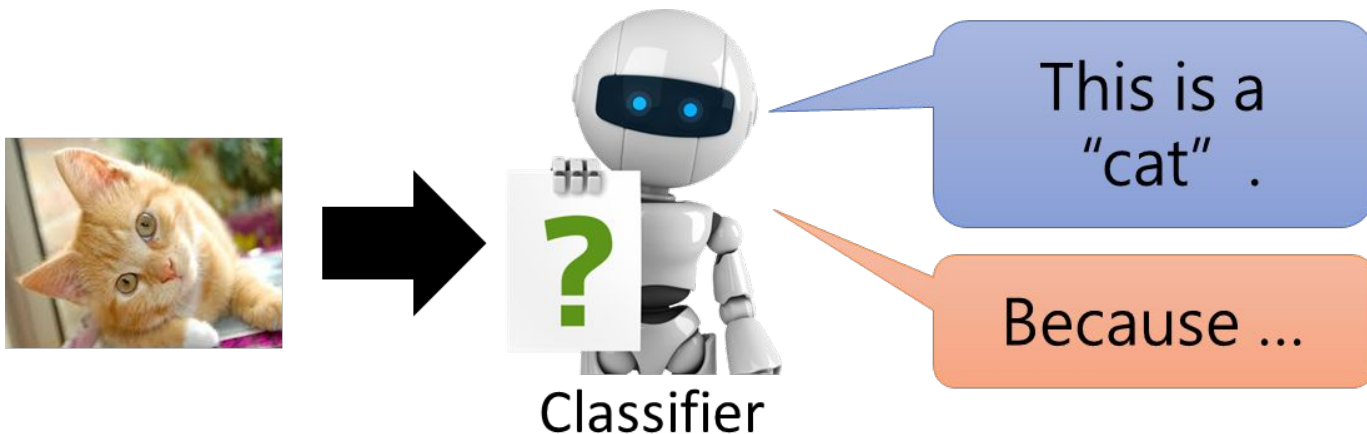
## *The Copy Machine Study* (Ellen Langer, Harvard University)

“Excuse me, I have 5 pages. May I use the Xerox machine?”  
60% accept

“Excuse me, I have 5 pages. May I use the Xerox machine,  
**because I’m in a rush?**”  
94% accept

“Excuse me, I have 5 pages. May I use the Xerox machine,  
**because I have to make copies?**”  
93% accept

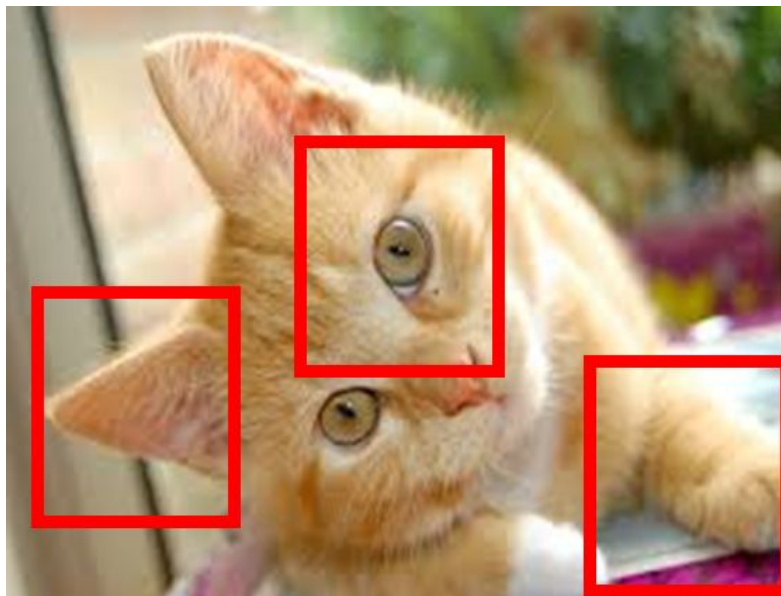
# Explainable AI



- Local Explanation
  - Why does the classifier think this image is a cat?
- Global Explanation
  - What does a "cat" look like? (not referred to a specific image)

# Local Explanation

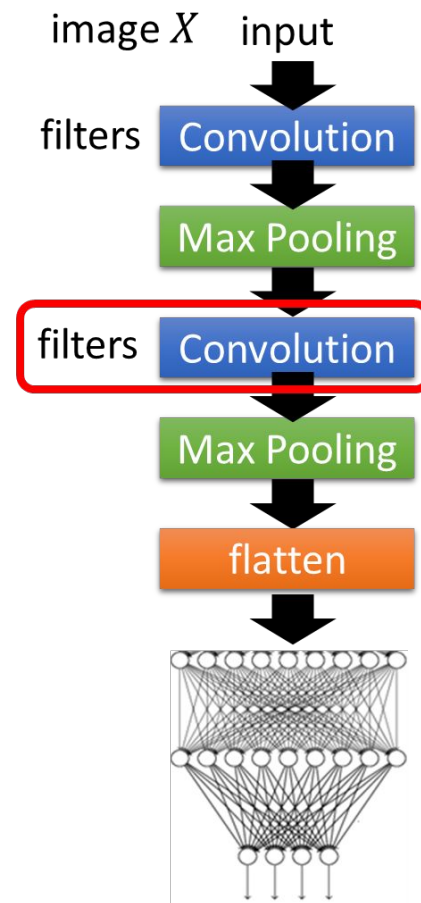
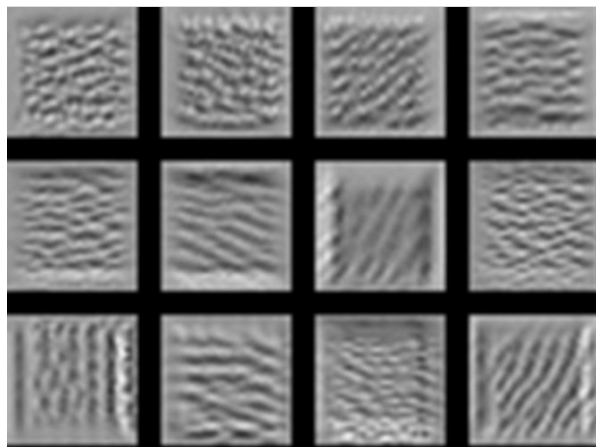
- Which component is critical for making decision?



# Global Explanation

- What does a filter detect?

$X^*$  for each filter



# Topic I: CNN explanation (hw3)



# Model: Food Classification

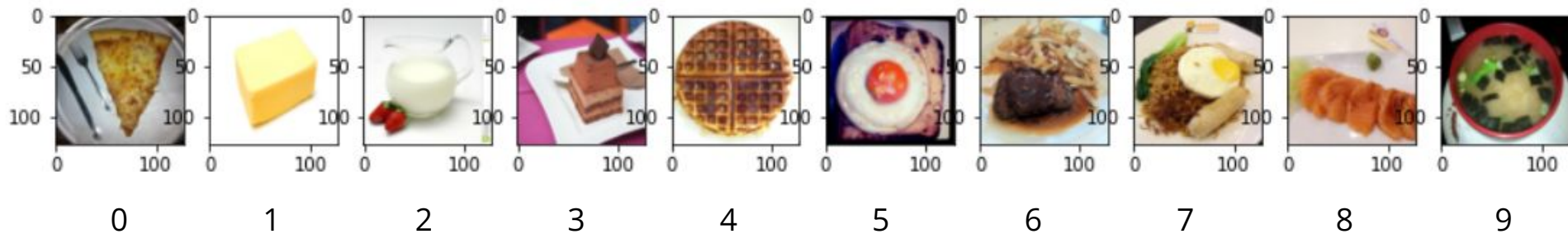
- We use a trained classifier model to do some explanations
- Model Structure: CNN model
- Dataset: 11 categories of food (same dataset in HW3)
  - Bread, Dairy product, Dessert, Egg, Fried food, Meat, Noodles/Pasta, Rice, Seafood, Soup, and Vegetables/Fruit

# Task

- Run the sample code and finish 20 questions (all multiple choice form)
- We'll cover 5 explanation approaches
  - Lime package
  - Saliency map
  - Smooth Grad
  - Filter Visualization
  - Integrated Gradients
- You need to:
  - Know the basic idea of each method
  - Run the code and observe the results
  - For some cases, you may need to modify a small part of the code

# Task: Observation

- In this homework, you only need to observe these 10 images.
- Please make sure **you got these 10 images in your code.**
- In the questions, the images are marked from **0 to 9.**
- We encourage you to observe other images!



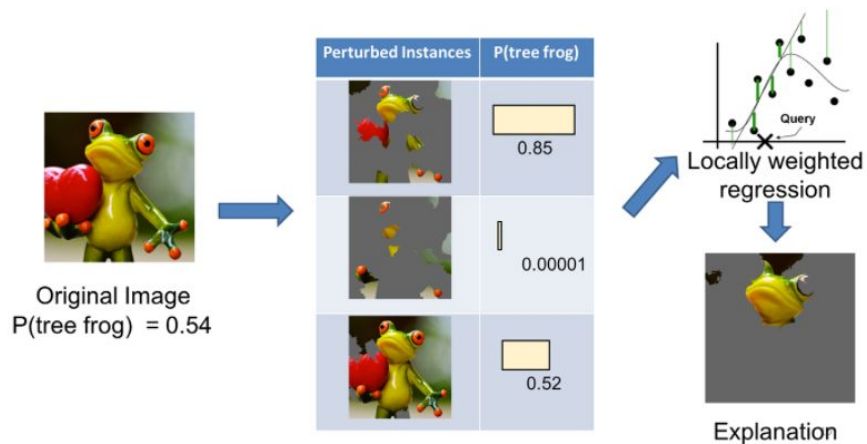
# Lime

## Question 1 to 4

- Install the Lime package -> `pip install lime==0.1.1.37`

GitHub repo: <https://github.com/marcotcr/lime>

Ref: <https://reurl.cc/5G8EGG>

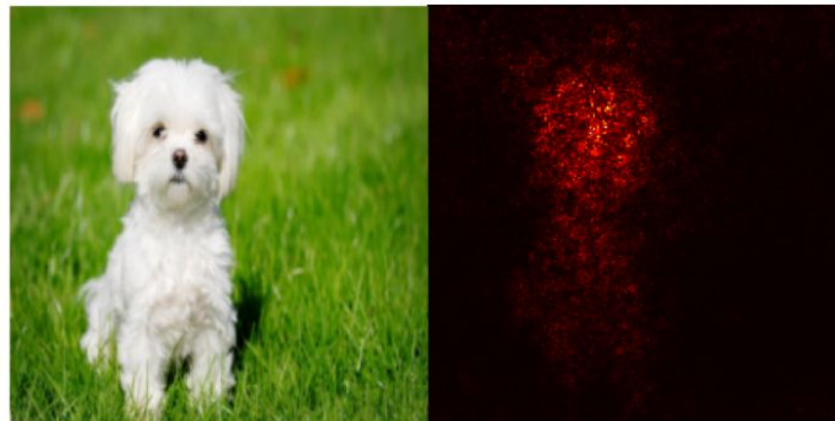


# Saliency Map

## Question 5 to 9

- Compute the gradient of output category with respect to input image.

Ref: <https://reurl.cc/6ELeLk>

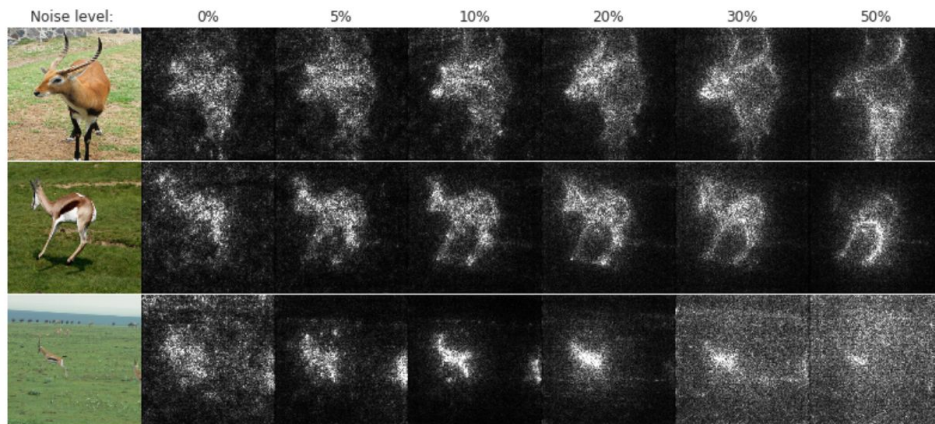


# Smooth Grad

## Question 10 to 13

- Randomly add noise to the input image, and get the heatmap. Just like what we did in the saliency method.

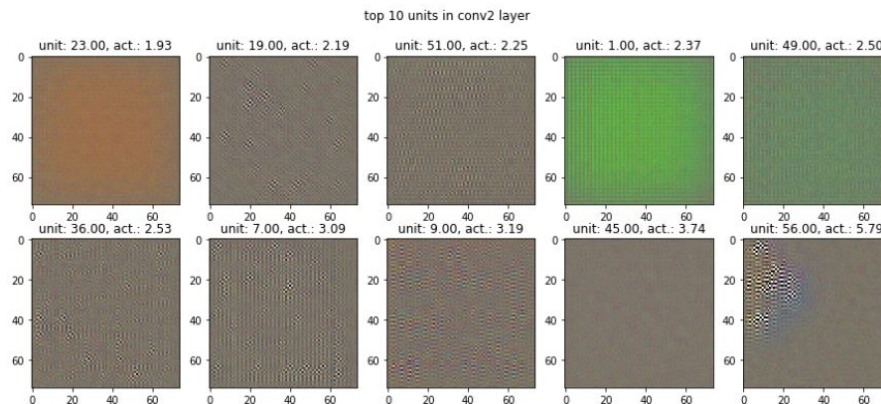
Ref: <https://arxiv.org/pdf/1706.03825.pdf>



# Filter Visualization

## Question 14 to 17

- Use Gradient Ascent method to find the image that activates the selected filter the most and plot them (start from white noise).



Ref: <https://reurl.cc/mGZNbA>

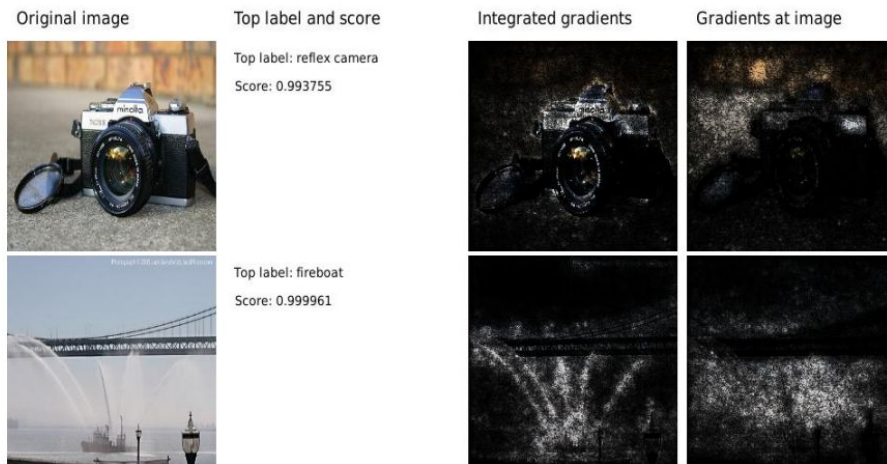
# Integrated Gradients

## Question 18 to 20

- Flexible baseline

$$\text{IntegratedGrads}_i(x) ::= (x_i - x'_i) \times \int_{\alpha=0}^1 \frac{\partial F(x' + \alpha \times (x - x'))}{\partial x_i} d\alpha$$

Ref: <https://arxiv.org/pdf/1703.01365.pdf>





# Topic II: BERT explanation (hw7)

# Task

- Run the sample code and finish 10 questions (all multiple choice form)
- We'll cover 3 explanation approaches
  - Attention Visualization
  - Embedding Visualization
  - Embedding analysis
- You need to:
  - Know the basic idea of each method
  - Run the code and observe the results
  - For some cases, you may need to modify a small part of the code

# Attention Visualization

## Question 21 to 24

- Visualize attention mechanism of bert using <https://exbert.net/exBERT.html>

Alternative link:

<https://huggingface.co/exbert/>

Ref: <https://arxiv.org/pdf/1910.05276.pdf>

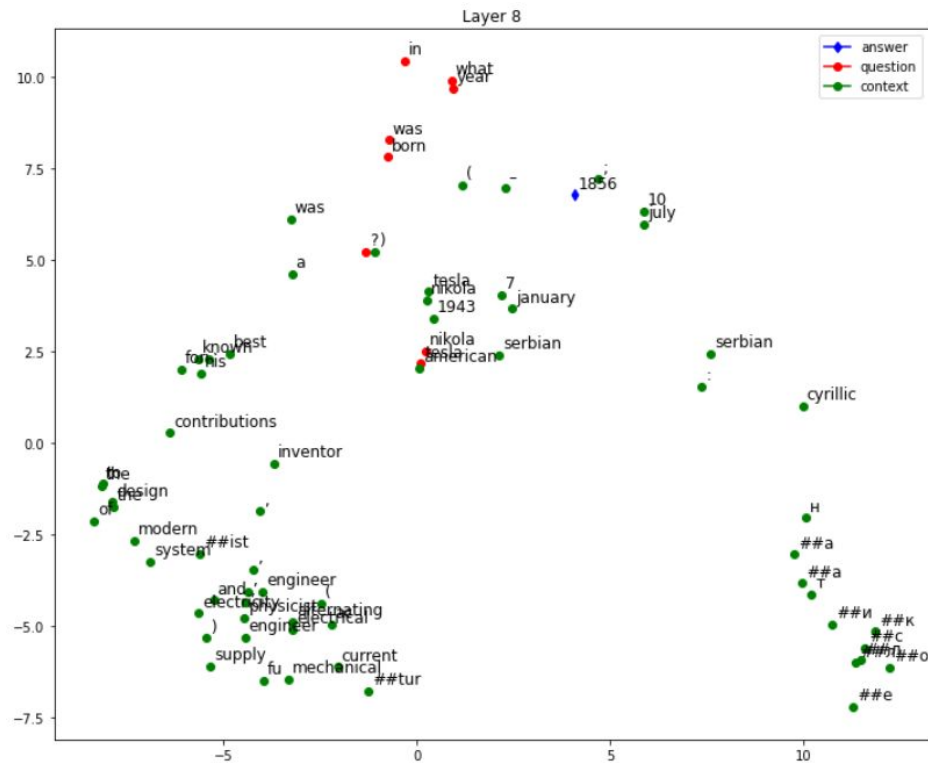
Tutorial: [https://youtu.be/e31oyfo\\_thY](https://youtu.be/e31oyfo_thY)

The screenshot displays the exBERT web application interface. At the top, the input sentence is "The girl ran to a local pub to escape the din of her city." Below this, the model is set to "bert-base-cased" and the attention is set to "Display top 70% of attention". The layer selector is set to "Layer: 1". The "Hide Special Tokens" toggle is turned on. The "Selected heads" are listed as "1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12". The main visualization is a bipartite graph showing attention weights between tokens on the left and right. The tokens on the left are "[CLS]", "The", "girl", "ran", "to", "a", "local", "pub", "to", "escape", "the", "din", "of", "her", "city", and "[SEP]". The tokens on the right are "[CLS]", "The", "girl", "ran", "to", "a", "local", "pub", "to", "escape", "the", "din", "of", "her", "city", and "[SEP]". The graph shows strong connections between the tokens on the left and the corresponding tokens on the right. A heatmap on the right side of the graph shows attention weights, with a prominent blue square indicating high attention for the "escape" token on the right.

# Embedding Visualization

## Question 25 to 27

- Visualize embedding across layers of BERT using PCA (Principal Component Analysis)
- Fine-tuned for Question Answering

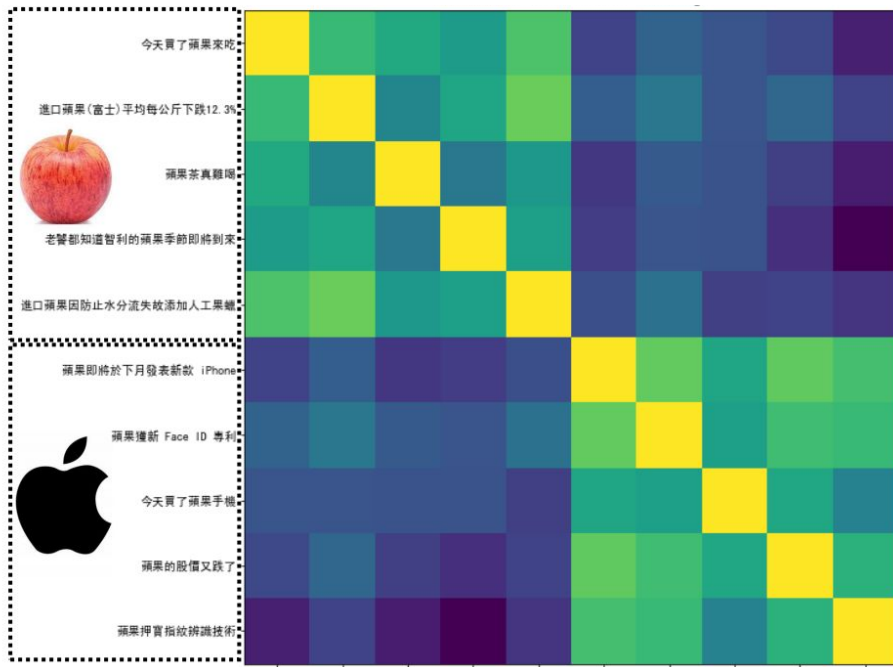


# Embedding Analysis

## Question 28 to 30

- Compare output embedding of BERT using:
  - Euclidean distance
  - Cosine similarity

You only need to change code in the section "TODO" !



# Grading

- 30 multiple choice questions
- CNN: 20 questions
  - 0.3 pt for each question
- BERT: 10 questions
  - 0.4 pt for each question
- You have to choose ALL the correct answers for each question
- No leaderboards & reports are needed!!

# Submission

- The questions are on gradescope
- Running the code may need some time!
- **No late submission!**
- You can answer the questions unlimited times
- The length of answering time of the assignment is unlimited
- We will consider the latest submission as the final submission
- **Remember to save the answer when answering the questions!**
- You will see the scores after the deadline only!
- Deadline: **2023/05/26 23:59 (UTC+8)**

# Links

- Code: [\[Colab\]](#)
- Gradescope: [\[Gradescope\]](#)

**Please don't change the original code,  
unless the question request you to do so.**



# If any questions, you can ask us via...

- NTU COOL (recommended)
  - [Homework 9 討論區](#)
- Email
  - [mlta-2023-spring@googlegroups.com](mailto:mlta-2023-spring@googlegroups.com)
  - The title **must** begin with “[hw9]”
- TA hours
  - In-person: Each Friday during class
  - Online: Each Monday night on google meet
  - Link: [HW9 TA hour](#)
  - 19:00 - 20:00 (Mandarin)
  - 20:00 - 21:00 (English)