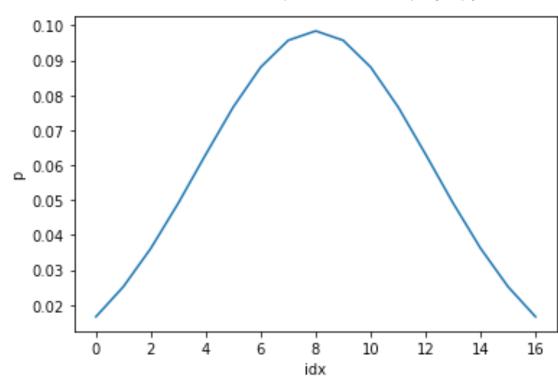
# CS 6476 Project 1

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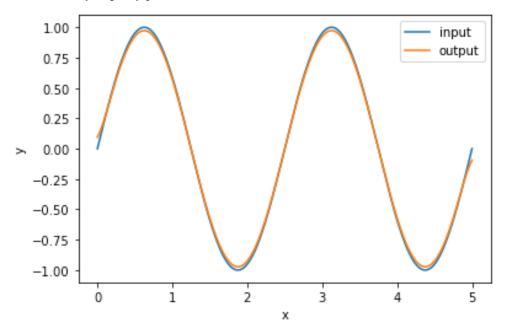
### Part 1: 1D Filter

<insert visualization of the low-pass filter from proj1.ipynb here>



### Part 1: 1D Filter

<insert visualization of filtered combined signal
from proj1.ipynb here>



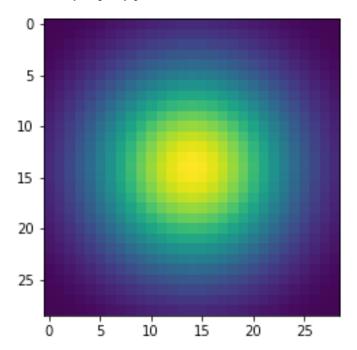
Describe your implementation in words and reflect on the checkpoint questions.

I create a 1D Gaussian low-pass filter to get the low frequency signal. And this filter also attenuate the high frequency while the low frequency is not affected.

Yes, the unit test passed.

# Part 2: Image Filtering

<insert visualization of the 2D Gaussian kernel
from proj1.ipynb here>



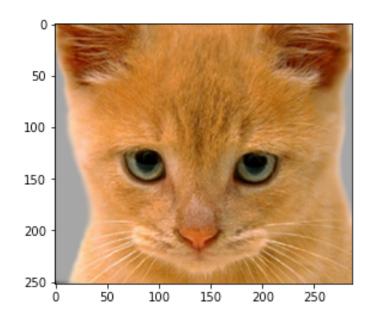
<Describe your implementation of my\_imfilter()
in words.>

I get the shape of image and filter first and use filter tensor to pad image tensor. Then use three for loops to create a new filtered image by calculating the dot product into tensor.

# Part 2: Image filtering

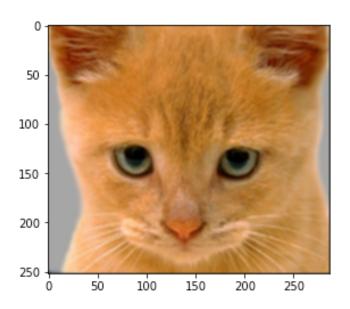
### **Identity filter**

<insert the results from proj1\_test\_filtering.ipynb
using 1b cat.bmp with the identity filter here>



#### Small blur with a box filter

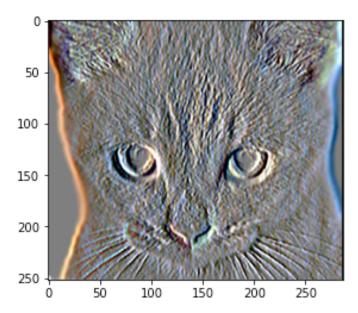
<insert the results from proj1\_test\_filtering.ipynb
using 1b\_cat.bmp with the box filter here>



# Part 2: Image filtering

### Sobel filter

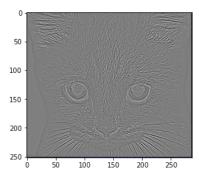
<insert the results from proj1\_test\_filtering.ipynb
using 1b\_cat.bmp with the Sobel filter here>

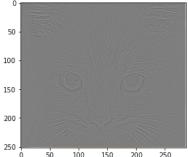


### **Discrete Laplacian filter**

<insert the results from proj1\_test\_filtering.ipynb
using 1b\_cat.bmp with the discrete Laplacian</pre>

filter here>





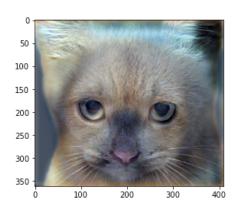
# Part 2: Hybrid images manually using Pytorch

<Describe your implementation of
create\_hybrid\_image() here.>

- 1. Dog x filter = low\_freq\_dog
- 2. Cat x filter = low\_freq\_cat
- 3. High\_freq\_cat = cat low\_freq\_cat
- 4. (low\_freq\_dog + low\_freq\_dog) / 2 = tmp
- 5. hybrid = torch.clamp(tmp, min = 0, max = 1.0)

### Cat + Dog

<insert your hybrid image here>



Cutoff frequency: <insert the value you used for this image pair> **7** 

# Part 2: Hybrid images manually using Pytorch

**Motorcycle + Bicycle** 

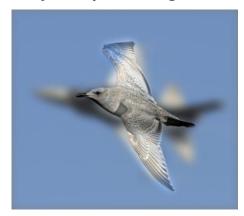
<insert your hybrid image here>



Cutoff frequency: <insert the value you used for this image pair> **7** 

Plane + Bird

<insert your hybrid image here>



Cutoff frequency: <insert the value you used for this image pair> **7** 

### Part 2: Hybrid images manually using Pytorch

### **Einstein + Marilyn**

<insert your hybrid image here>



Cutoff frequency: <insert the value you used for this image pair> **7** 

### **Submarine + Fish**

<insert your hybrid image here>



Cutoff frequency: <insert the value you used for this image pair> **7** 

# Part 3: Hybrid images with PyTorch operators

Cat + Dog

<insert your hybrid image here>



**Motorcycle + Bicycle** 

<insert your hybrid image here>



# Part 3: Hybrid images with PyTorch operators

Plane + Bird

<insert your hybrid image here>



**Einstein + Marilyn** 

<insert your hybrid image here>



### Part 3: Hybrid images with PyTorch operators

#### Submarine + Fish

<insert your hybrid image here>



#### Part 1 vs. Part 2

<Compare the run-times of Parts 2 and 3 here, as calculated in proj1.ipynb. What can you say about the two methods?>

Part2: 45-55 sec

Part3: 0.18-0.25 sec

Torch.nn.functional.conv2d is much faster but when I implement myself, I can totally understand what inside the conv2d API

### **Tests**

<Provide a screenshot of the results when you run `pytest tests` on your final code implementation (note: we will re-run these tests).>

### Conclusions

<Describe what you have learned in this project. Consider questions like how varying the cutoff standard deviation value or swapping images within a pair influences the resulting hybrid image. Feel free to include any challenges you ran into.>

# Extra Credit

# Image Filtering using DFT

<insert visualization of the DFT filtered
6a\_dog.bmp from proj1.ipynb here>

Describe your implementation in words.

# Add some cool hybrid images!