

# CS 445: Computational Photography

## Programming Project #3: Gradient Domain Fusion

In [2]:

```
# from google.colab import drive
# drive.mount('/content/drive')
```

In [1]:

```
import cv2
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import os
from random import random
import time
import scipy
import scipy as sc
import scipy.sparse.linalg
from scipy.sparse import *
from scipy.sparse.linalg import lsqr
from utils import *

# modify to where you store your project data including utils.py
# datadir = "/content/drive/My Drive/cs445_projects/proj3/"

# utilfn = datadir + "utils.py"
# !cp "$utilfn" .
# samplesfn = datadir + "samples"
# !cp -r "$samplesfn" .
import utils
```

## Part 1 Toy Problem (20 pts)

In [2]:

```
def toy_reconstruct(img):
    """
    The implementation for gradient domain processing is not complicated, but it
    is easy to make a mistake, so let's start with a toy example. Reconstruct this i
    mage from its gradient values, plus one pixel intensity. Denote the intensity of
    the source image at (x, y) as s(x,y) and the value to solve for as v(x,y). For e
    ach pixel, then, we have two objectives:
    1. minimize (v(x+1,y)-v(x,y) - (s(x+1,y)-s(x,y)))^2
    2. minimize (v(x,y+1)-v(x,y) - (s(x,y+1)-s(x,y)))^2
    Note that these could be solved while adding any constant value to v, so we
    will add one more objective:
    3. minimize (v(1,1)-s(1,1))^2
    :param toy_img: numpy.ndarray
    """

    img_h = img.shape[0]
    img_w = img.shape[1]
    #create a np array with shape of 1 * (img_h * img_w)
    im2var = np.arange(img_h * img_w)
    #reshape the single row array into the shape we want
    im2var = im2var.reshape(img_h, img_w)
#    print(im2var)
#    we use im2var to locate the pixel in original img

#    h = img_h * (img_w - 1) + (img_h - 1) * img_w + 1
h = img_h * img_w * 2 #26180
#    print("h" , h) #h 25952

w = img_h * img_w
#    print("w" , w) #w 13090

#    A has size of :
#h: # of formulas/constraints
#w: # of values we want to resolve, which is the # of pixels in our target l
ocation, which is img
#check 5.1.3 ~13mins for more details
#    A = np.zeros([h,w])
A = scipy.sparse.lil_matrix((h, w), dtype='double')
b = np.zeros((A.shape[0],1), dtype = 'double')
e = 0

#    A[e, im2var[y][x]] not A[e][im2var[y][x]]
A[e, im2var[0][0]] = 1
b[e] = img[0][0]
e = e+1
A[e, im2var[0][0]] = 1
b[e] = img[0][0]

print(im2var.shape)
for y in range(img_h - 1):
    for x in range(img_w - 1):
        #copy from tips, obj1
        A[e, im2var[y][x+1]] = 1 #the coefficient for variables v, and b is
the constant from original image
        A[e, im2var[y][x]] = -1
        b[e] = img[y][x+1] - img[y][x]
        e = e + 1
        #obj2
```

```

        A[e, im2var[y+1][x]] = 1
        A[e, im2var[y][x]] = -1
        b[e] = img[y+1][x] - img[y][x]
        e = e + 1
A[e, im2var[y+1][x+1]] = 1
b[e] = img[y+1][x+1]

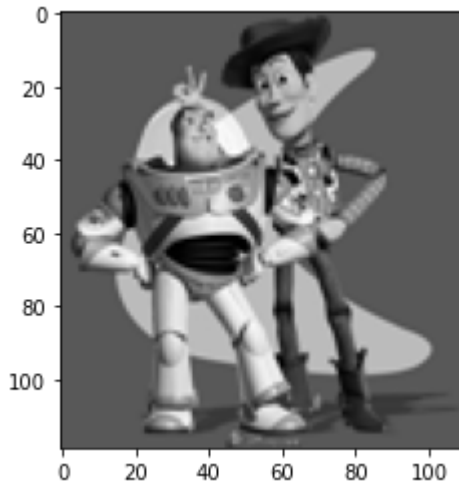
print("A.shape",A.shape)
print("b.shape", b.shape)
#     A_spa = csr_matrix(A)
#     v = lsqr(A_spa, b)
v = scipy.sparse.linalg.lsqr(A.tocsr(), b);
#https://stackoverflow.com/questions/48621407/python-equivalent-of-matlabs-lsqr-
#https://het.as.utexas.edu/HET/Software/Scipy/generated/scipy.sparse.linalg.lsqr.html
#     v = scipy.sparse.linalg.lsqr(A.tocsr(), b)
#     print("A_csr.shape",A_csr.shape)
#     print("A.shape",A.shape)
#     print(v)
return v[0].reshape(img_h, img_w)
#     return img

```

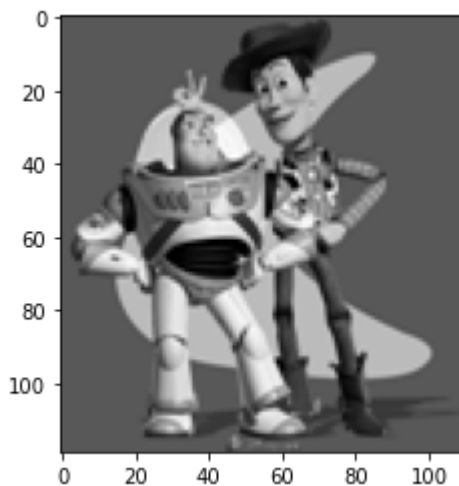
In [3]:

```
toy_img = cv2.cvtColor(cv2.imread('samples/toy_problem.png'), cv2.COLOR_BGR2GRAY)
).astype('double') / 255.0
plt.imshow(toy_img, cmap="gray")
plt.show()

im_out = toy_reconstruct(toy_img)
plt.imshow(im_out, cmap="gray")
plt.show()
print("Max error is: ", np.sqrt(((im_out - toy_img)**2).max()))
```



```
(119, 110)
A.shape (26180, 13090)
b.shape (26180, 1)
```



Max error is: 8.932118298277114e-06

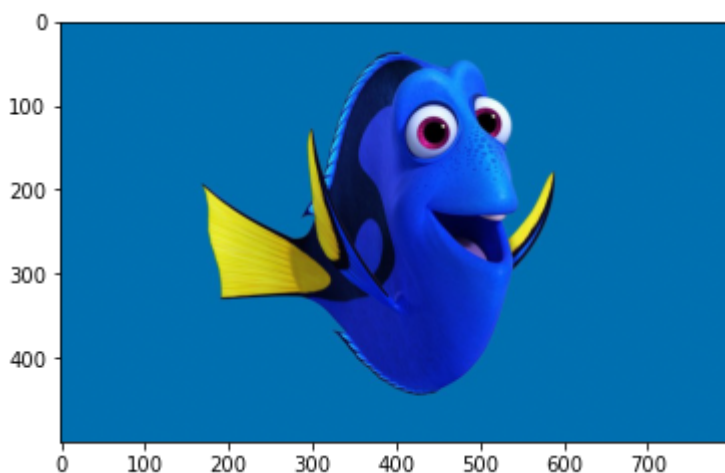
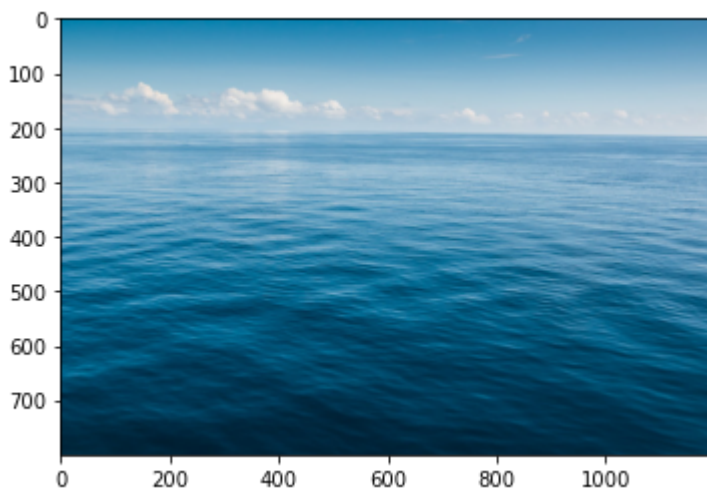
## Preparation

In [4]:

```
background_img = cv2.cvtColor(cv2.imread('samples/comb1_1.jpeg'), cv2.COLOR_BGR2RGB).astype('double') / 255.0
plt.figure()
plt.imshow(background_img)
plt.show()
object_img = cv2.cvtColor(cv2.imread('samples/comb1_2.jpeg'), cv2.COLOR_BGR2RGB).astype('double') / 255.0
plt.imshow(object_img)
plt.show()

use_interface = True # set to true if you want to use the interface to choose points (might not work in Colab)
if not use_interface:
    xs = (65, 359, 359, 65)
    ys = (24, 24, 457, 457)
    object_mask = utils.get_mask(ys, xs, object_img)
    bottom_center = (500, 2500) # (x,y)

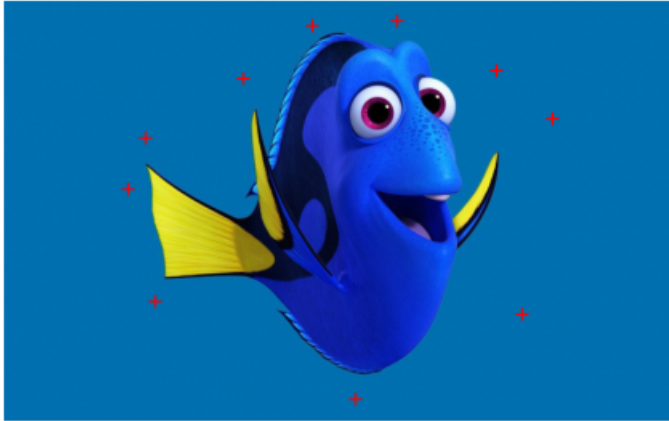
    object_img, object_mask = utils.crop_object_img(object_img, object_mask)
    bg_ul = utils.upper_left_background_rc(object_mask, bottom_center)
    plt.imshow(utils.get_combined_img(background_img, object_img, object_mask, bg_ul))
```



In [6]:

```
if use_interface:
    import matplotlib.pyplot as plt
    %matplotlib notebook
    mask_coords = specify_mask(object_img)
```

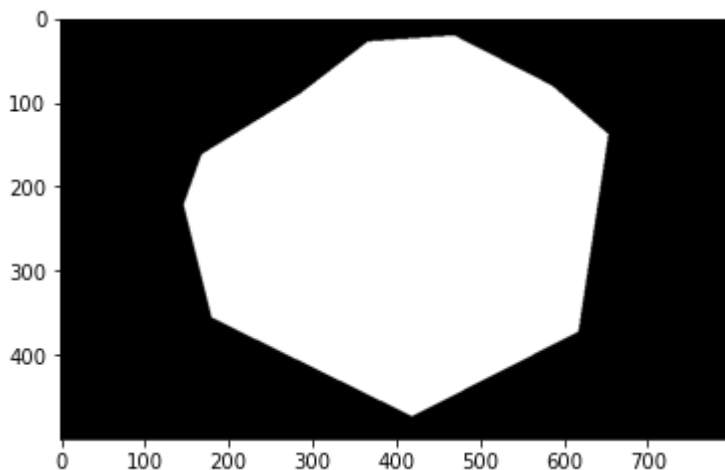
If it doesn't get you to the drawing mode, then rerun this function again.



In [7]:

```
if use_interface:
    xs = mask_coords[0]
    ys = mask_coords[1]
    %matplotlib inline
    import matplotlib.pyplot as plt
    plt.figure()
    object_mask = get_mask(ys, xs, object_img)
```

<Figure size 432x288 with 0 Axes>



In [8]:

```
if use_interface:
    %matplotlib notebook
    import matplotlib.pyplot as plt
    bottom_center = specify_bottom_center(background_img)
    #     bottom_center = background_img.shape[0],background_img.shape[1]
```

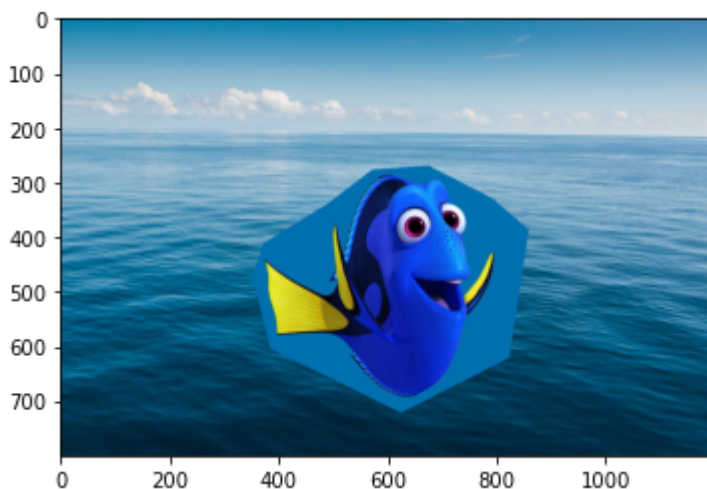
If it doesn't get you to the drawing mode, then rerun this function again. Also, make sure the object fill fit into the background image. Otherwise it will crash



In [9]:

```
# print(bottom_center)
if use_interface:
    %matplotlib inline
    import matplotlib.pyplot as plt

    object_img, object_mask = utils.crop_object_img(object_img, object_mask)
    bg_ul = utils.upper_left_background_rc(object_mask, bottom_center)
    plt.imshow(utils.get_combined_img(background_img, object_img, object_mask, b
g_ul))
#     plt.savefig('comb3_direct.jpg')
```

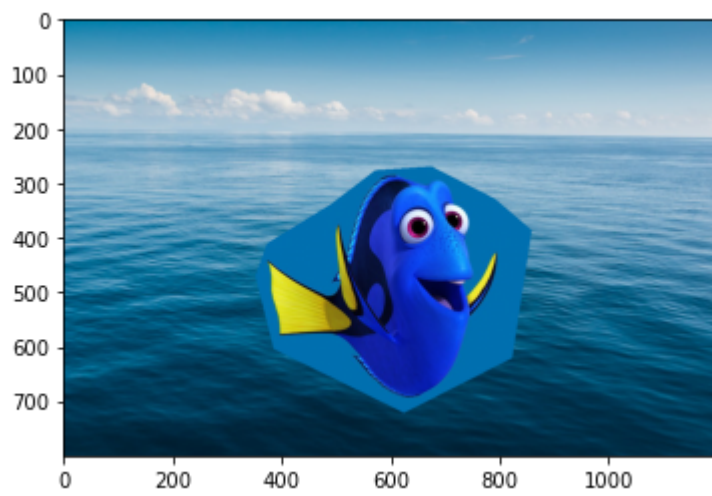


## **Part 2 Poisson Blending (50 pts)**

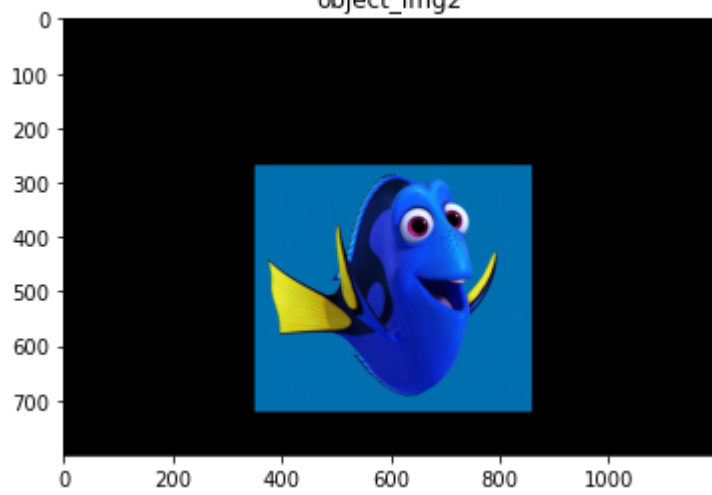


In [10]:

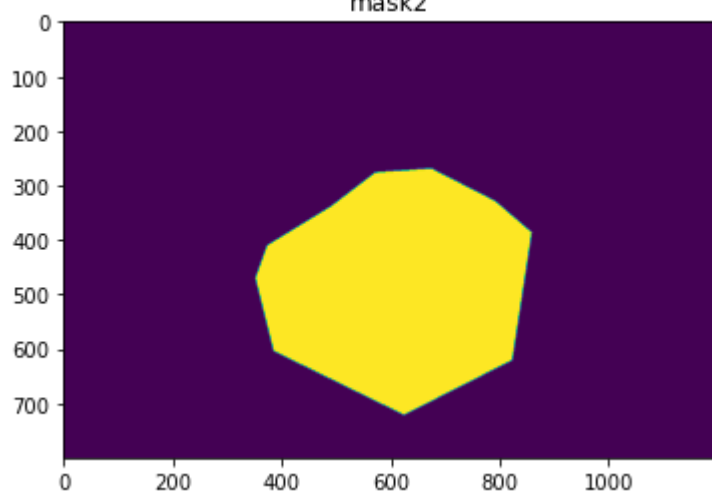
```
object_img2, mask = align_source(object_img, object_mask, background_img, bottom_center)
object_img3, mask2 = align_source(object_img, mask, background_img, bottom_center)
plt.imshow(object_img2)
plt.title('object_img2')
plt.show()
plt.imshow(mask2)
plt.title('mask2')
plt.show()
```



object\_img2

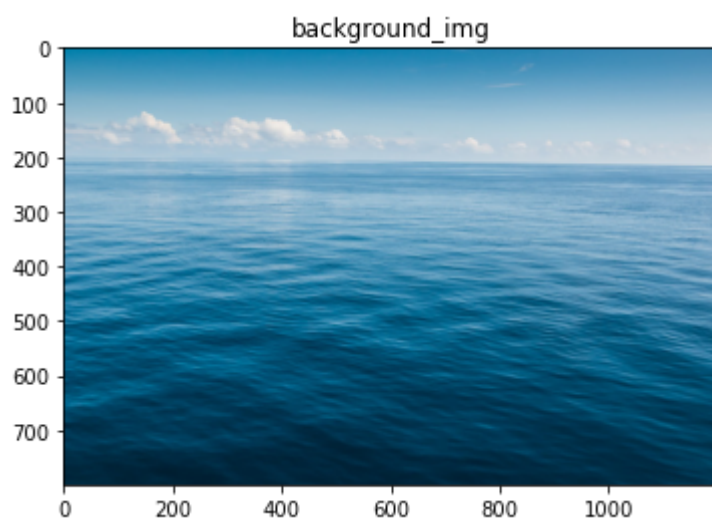
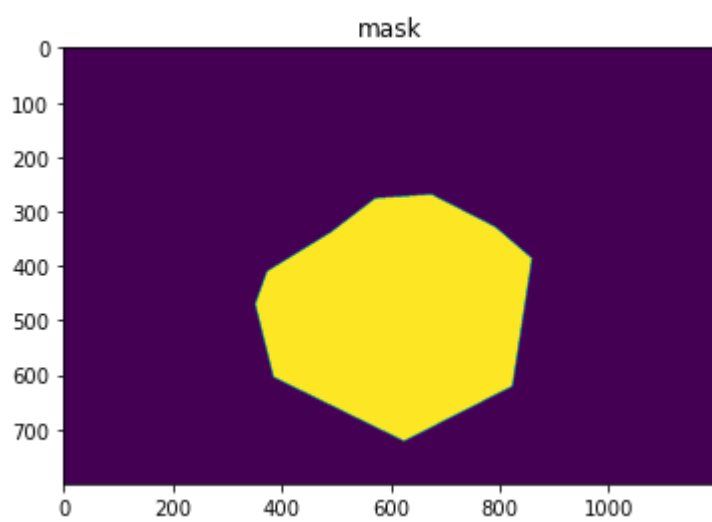
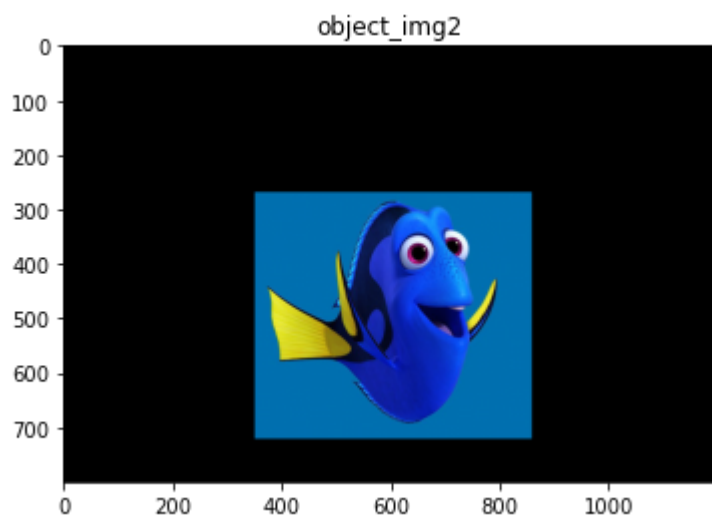


mask2



In [11]:

```
plt.imshow(object_img2)
plt.title("object_img2")
plt.show()
plt.imshow(mask)
plt.title("mask")
plt.show()
plt.imshow(background_img)
plt.title("background_img")
plt.show()
```



In [12]:

```
def poisson_blend(object_img2, mask2, bg_img, bg_ul):
    """
    Returns a Poisson blended image with masked object_img over the bg_img at po
    sition specified by bg_ul.
    Can be implemented to operate on a single channel or multiple channels
    :param object_img: the image containing the foreground object
    :param object_mask: the mask of the foreground object in object_img
    :param background_img: the background image
    :param bg_ul: position (row, col) in background image corresponding to (0,0)
    of object_img
    reference:
    https://cs.brown.edu/courses/csci1950-g/results/proj2/edwallac/
    http://cs.brown.edu/courses/csl29/results/proj2/taox/

    """
    non_zero_x, non_zero_y = np.where(mask2 == 1)

    img_h = object_img2.shape[0]
    img_w = object_img2.shape[1]
    im2var = np.arange(img_h*img_w).reshape(img_h, img_w)
    result = bg_img
    e = 0
    # h = img_h * img_w * 4 # is this varibale setup right? does it matter?
    h = len(non_zero_x) * 4 # is this varibale setup right? does it matter?
    print(h)
    w = img_h * img_w
    # A = np.zeros((h,w), dtype = np.float64)
    # A = scipy.sparse.lil_matrix((neq, nr*nc), dtype='double') # init lil
    A = scipy.sparse.lil_matrix((h,w),dtype=np.float64)

    b = np.zeros((A.shape[0],1), dtype = np.float64)
    # print(im2var.shape)

    print("checkpoint 1")
    # A[e, im2var[y][x]] not A[e][im2var[y][x]]
    for y in range(img_h):
        for x in range(img_w):
            if mask2[y][x] == 0:
                continue
            else:
                result[y][x] = 0
                if mask2[y][x+1] == 1:
                    A[e, im2var[y][x+1]] = -1
                    A[e, im2var[y][x]] = 1
                    b[e] = object_img2[y][x] - object_img2[y][x+1]
                    e = e + 1
                else:
                    A[e, im2var[y][x]] = 1
                    b[e] = object_img2[y][x] - object_img2[y][x+1] + bg_img[y][x
+1]
                    e = e + 1
#####
                if mask2[y][x-1] == 1:
                    A[e, im2var[y][x-1]] = -1
                    A[e, im2var[y][x]] = 1
                    b[e] = object_img2[y][x] - object_img2[y][x-1]
                    e = e + 1
                else:
```

```

        A[e, im2var[y][x]] = 1
        b[e] = object_img2[y][x] - object_img2[y][x-1] + bg_img[y][x
-1]

        e = e + 1
#####
        if mask2[y+1][x] == 1:
            A[e, im2var[y+1][x]] = -1
            A[e, im2var[y][x]] = 1
            b[e] = object_img2[y][x] - object_img2[y+1][x]
            e = e + 1
        else:
            A[e, im2var[y][x]] = 1
            b[e] = object_img2[y][x] - object_img2[y+1][x] + bg_img[y+1]
[x]

            e = e + 1
#####
            if mask2[y-1][x] == 1:
                A[e, im2var[y-1][x]] = -1
                A[e, im2var[y][x]] = 1
                b[e] = object_img2[y][x] - object_img2[y-1][x]
                e = e + 1
            else:
                A[e, im2var[y][x]] = 1
                b[e] = object_img2[y][x] - object_img2[y-1][x] + bg_img[y-1]
[x]

            e = e + 1
    print("calculating V ")
#     print(A.shape)
#     print(b.shape)

#computer dies while runing the A_spa and calculate v
#     A_spa = csr_matrix(A)
#     v = lsqr(A_spa, b)
v = scipy.sparse.linalg.lsqr(A.tocsr(), b);
print("v calculate complete")
#     print("len(v)", len(v))
#     print("len(v[0])", len(v[0]))

    result = result + v[0].reshape(img_h, img_w)

#     return v[0].reshape(img_h, img_w)
return result

```

In [13]:

```
im_blend = np.zeros(background_img.shape)
# plt.imshow(mask2)
# plt.title('mask2')
# plt.show()
# plt.imshow(background_img)
# plt.title('background_img')
# plt.show()
# plt.imshow(object_img2)
# plt.title('object_img2')
# plt.show()

for b in np.arange(3):
    im_blend[:, :, b] = poisson_blend(object_img2[:, :, b], mask, background_img
[:, :, b].copy(), bg_ul)

plt.figure(figsize=(10,10))
plt.imshow(im_blend)

# plt.savefig('comb3_poi.jpg')
```

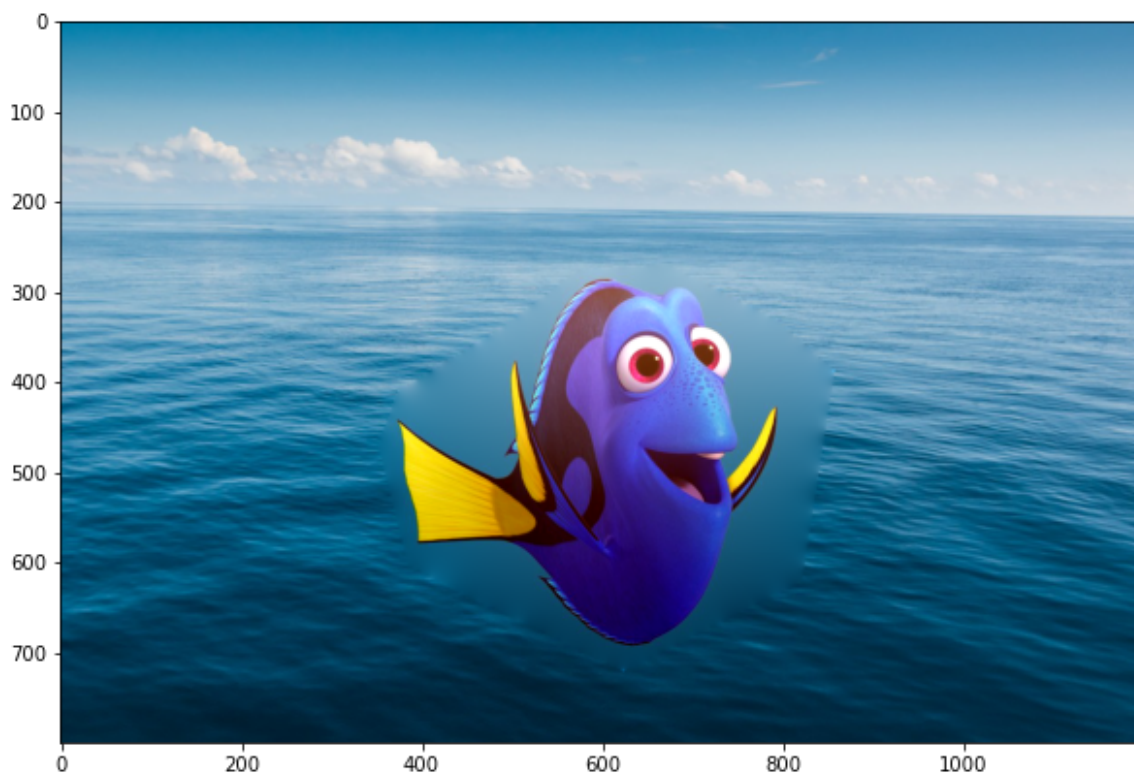
```
648004
checkpoint 1
calculating V
v calculate complete
648004
checkpoint 1
calculating V
v calculate complete
648004
checkpoint 1
calculating V
```

Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

```
v calculate complete
```

Out[13]:

<matplotlib.image.AxesImage at 0xa202519b0>



## Part 3 Mixed Gradients (20 pts)

In [15]:

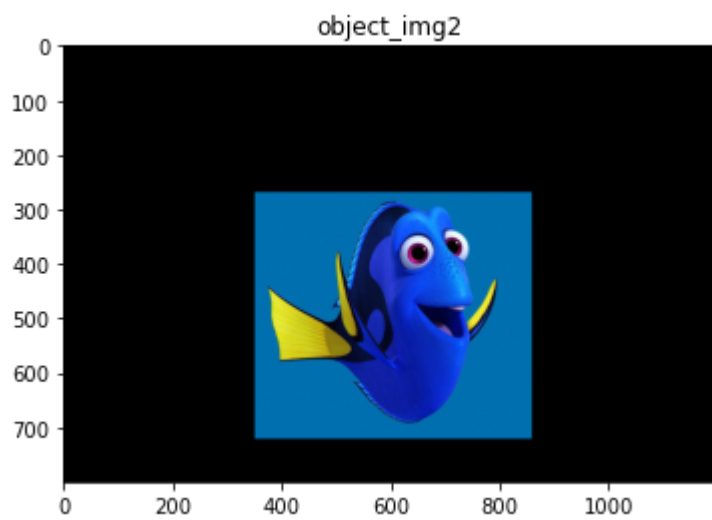
```
def greater(input1, input2):
    if abs(input1) > abs(input2):
        return input1
    else:
        return input2
print(greater(-20, 9))
```

-20



In [16]:

```
plt.imshow(object_img2)  
plt.title('object_img2')  
plt.show()
```



In [17]:

```
def mixed_blend(object_img2, mask2, bg_img, bg_ul):
    """
    Returns a mixed gradient blended image with masked object_img over the bg_img
    at position specified by bg_ul.
    Can be implemented to operate on a single channel or multiple channels
    :param object_img: the image containing the foreground object
    :param object_mask: the mask of the foreground object in object_img
    :param background_img: the background image
    :param bg_ul: position (row, col) in background image corresponding to (0,0)
    of object_img
    """
    non_zero_x, non_zero_y = np.where(mask2 == 1)

    img_h = object_img2.shape[0]
    img_w = object_img2.shape[1]
    im2var = np.arange(img_h*img_w).reshape(img_h, img_w)
    # result = bg_img
    result = np.zeros((bg_img.shape), dtype='double')
    result = (1-mask2) * bg_img

    e = 0
    h = len(non_zero_x) * 4

    w = img_h * img_w
    A = scipy.sparse.lil_matrix((h,w), dtype=np.float64)
    b = np.zeros((A.shape[0],1), dtype = np.float64)

    print("checkpoint 1")
    for y in range(img_h):
        for x in range(img_w):
            if mask2[y][x] == 0:
                continue
            else:
                # result[y][x] = 0

                if mask2[y][x+1] == 1:
                    A[e, im2var[y][x+1]] = -1
                    A[e, im2var[y][x]] = 1
                    # b[e] = object_img2[y][x] - object_img2[y][x+1]
                    b[e] = compare(object_img2[y][x] - object_img2[y][x+1], bg_img[y][x] - bg_img[y][x+1])
                    e = e + 1
                else:
                    A[e, im2var[y][x]] = 1
                    # b[e] = object_img2[y][x] - object_img2[y][x+1] + bg_img[y][x+1]
                    b[e] = compare(object_img2[y][x] - object_img2[y][x+1], bg_img[y][x] - bg_img[y][x+1]) + bg_img[y][x+1]
                    e = e + 1
                #####
                if mask2[y][x-1] == 1:
                    A[e, im2var[y][x-1]] = -1
                    A[e, im2var[y][x]] = 1
                    b[e] = object_img2[y][x] - object_img2[y][x-1]
                    # b[e] = compare(object_img2[y][x] - object_img2[y][x-1], bg_img[y][x] - bg_img[y][x-1])
                    e = e + 1
```

```

        else:
            A[e, im2var[y][x]] = 1
#            b[e] = object_img2[y][x] - object_img2[y][x-1] + bg_img[y]
#            [x-1]
            b[e] = compare(object_img2[y][x] - object_img2[y][x-1], bg_img[y][x] - bg_img[y][x-1]) + bg_img[y][x-1]
            e = e + 1
#####
        if mask2[y+1][x] == 1:
            A[e, im2var[y+1][x]] = -1
            A[e, im2var[y][x]] = 1

            b[e] = compare(object_img2[y][x] - object_img2[y+1][x], bg_img[y][x] - bg_img[y+1][x])
            e = e + 1
        else:
            A[e, im2var[y][x]] = 1
            b[e] = compare(object_img2[y][x] - object_img2[y+1][x], bg_img[y][x] - bg_img[y+1][x]) + bg_img[y+1][x]
            e = e + 1
#####
        if mask2[y-1][x] == 1:
            A[e, im2var[y-1][x]] = -1
            A[e, im2var[y][x]] = 1
            b[e] = compare(object_img2[y][x] - object_img2[y-1][x], bg_img[y][x] - bg_img[y-1][x])
            e = e + 1
        else:
            A[e, im2var[y][x]] = 1
            b[e] = compare(object_img2[y][x] - object_img2[y-1][x], bg_img[y][x] - bg_img[y-1][x]) + bg_img[y-1][x]
            e = e + 1
    print("calculating V ")
#    print(A.shape)
#    print(b.shape)

v = scipy.sparse.linalg.lsqr(A.tocsr(), b);
print("v calculate complete")
result = result + v[0].reshape(img_h, img_w)
return result

```

In [18]:

```
im_mix = np.zeros(background_img.shape)
for b in np.arange(3):
    im_mix[:, :, b] = mixed_blend(object_img2[:, :, b], mask, background_img[:, :, b].
copy(), bg_ul)

plt.figure(figsize=(10,10))
plt.imshow(im_mix)
```

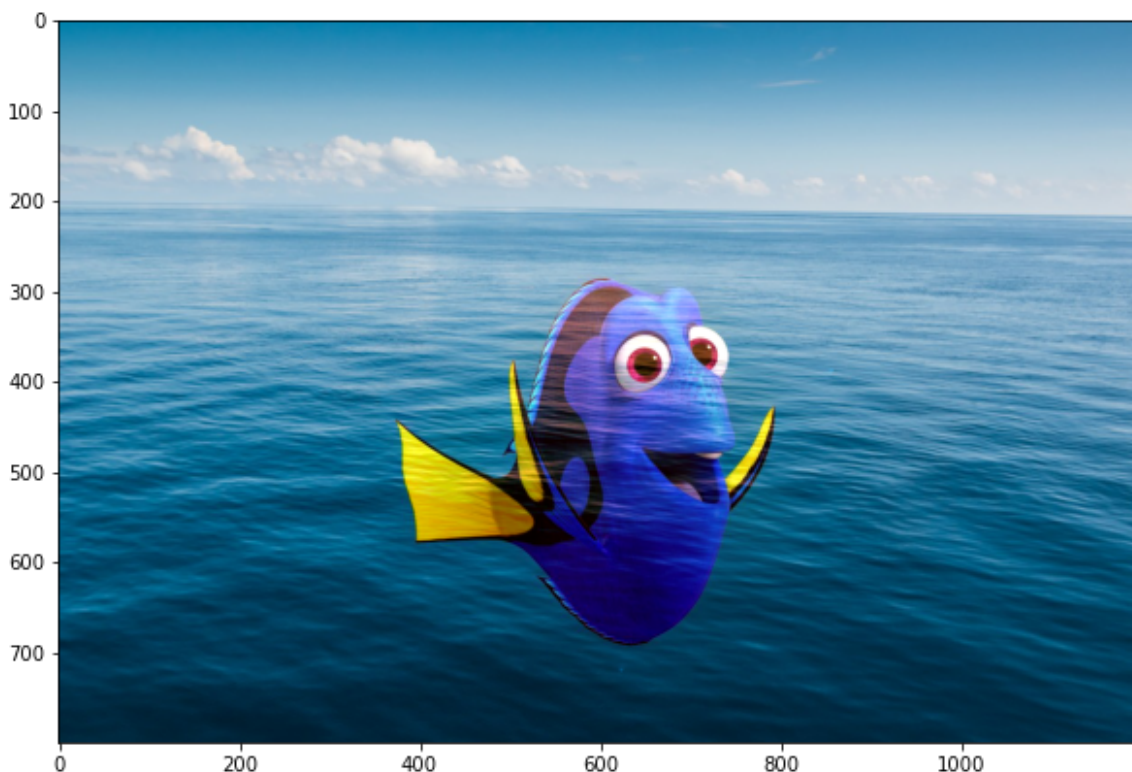
```
checkpoint 1
calculating V
v calculate complete
checkpoint 1
calculating V
v calculate complete
checkpoint 1
calculating V
```

Clipping input data to the valid range for imshow with RGB data  
([0..1] for floats or [0..255] for integers).

```
v calculate complete
```

Out[18]:

<matplotlib.image.AxesImage at 0xa2056ff60>



## Bells & Whistles (Extra Points)

### Color2Gray (20 pts)

In [ ]:

```
def color2gray(img):  
    pass
```

## Laplacian pyramid blending (20 pts)

In [ ]:

```
def laplacian_blend(object_img, object_mask, bg_img, bg_ul):  
    # feel free to change input parameters  
    pass
```

## More gradient domain processing (up to 20 pts)

In [ ]: