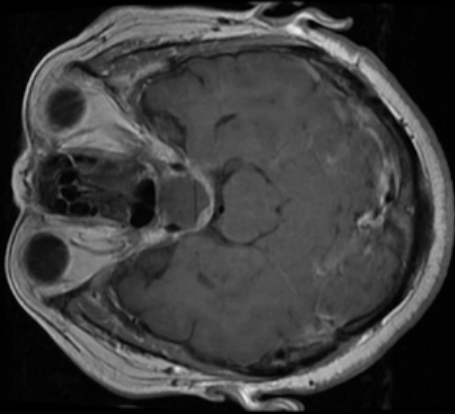


Brain Cancer Segmentation

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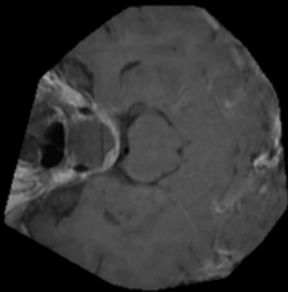
Segmentation

1. Skull Stripping
2. Mean color of the tumor
3. Segmentation of the tumor



Skull Stripping

The goal of this stage is to go and eliminate the edge of the brain and leave only the brain tissue



Mean color of the tumor

In this phase the average color of the brain is analyzed and areas of pixels are sought that have an average color quite distant from the color of the brain

```
...  
this function is used to segment the image using K-Means Clustering.  
...  
def KMeansClustering(data, k, img):  
    kmeans = KMeans(n_clusters=k, random_state=0)  
    kmeans.fit(data)  
    labels = kmeans.predict(data)  
    segmented_data = np.uint8(kmeans.cluster_centers_[labels])  
    segmented_img = segmented_data.reshape(img.shape)  
    return segmented_img
```

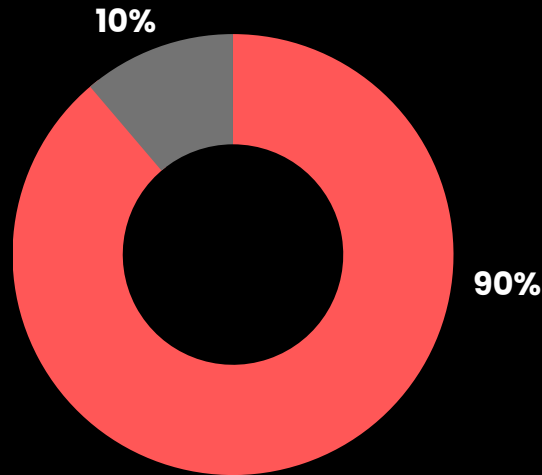
```
def get_tumor(color_tumor, brain, segm, mylist, area_contorno_esterno, colore_cervello):
    lista_contorni = []
    for value in mylist:
        mask = np.zeros(brain.shape[:2], np.uint8)
        mask[segm == value] = 255
        contours, hierarchy = cv2.findContours(mask, cv2.RETR_TREE, cv2.CHAIN_APPROX_SIMPLE)
        for el in contours:
            el = cv2.convexHull(el)
            area = cv2.contourArea(el)
            circumference = cv2.arcLength(el, True)
            if circumference == 0:
                continue
            circularity = 4 * np.pi * (area / (circumference * circumference))
            if (circularity >= 0.3 and area >= 0.005 * area_contorno_esterno and area <=
0.15 * area_contorno_esterno) or (circularity >= 0.45 and area > 0.15 * area_contorno_esterno and area <=
0.20 * area_contorno_esterno) or (circularity >= 0.55 and area > 0.20 * area_contorno_esterno and area
<= 0.22 * area_contorno_esterno):
                mask2 = np.zeros(brain.shape[:2], dtype="uint8")
                diz = {}
                cv2.drawContours(mask2, [el], -1, 255, -1)
                for i in range(0, brain.shape[0]):
                    for j in range(0, brain.shape[1]):
                        if mask2[i][j] == 255:
                            if brain[i][j] in diz:
                                diz[brain[i][j]] += 1
                            else:
                                diz[brain[i][j]] = 1
                somma = 0
                for k,v in diz.items():
                    somma += k
                media = somma/len(diz)
                lista_contorni.append((el, media, area, circularity))
```

```
diz_contorni = []
for tup in lista_contorni:
    media_pesata = get_media_pesata(tup, color_tumor, area_contorno_esterno, colore_cervello)
    diz_contorni.append(media_pesata)
brain = cv2.cvtColor(brain, cv2.COLOR_GRAY2BGR)
index = 0
diz_appoggio = {}
for el in lista_contorni:
    r_random = np.random.randint(0, 255)
    g_random = np.random.randint(0, 255)
    b_random = np.random.randint(0, 255)
    cl = (r_random, g_random, b_random)
    diz_appoggio[(diz_contorni[index], cl)] = (el[0], el[1])
    cv2.drawContours(brain, [el[0]], -1, cl, 2)
    index += 1
dizionario_2 = {}
for k,v in diz_appoggio.items():
    if k[0] >= 1.15:
        dizionario_2[k] = v
diz_appoggio = dizionario_2
diz_appoggio = dict(sorted(diz_appoggio.items(), key=lambda item: item[0][0]))
diz_appoggio = dict(reversed(list(diz_appoggio.items())))
if len(diz_appoggio) == 0:
    return "no tumor", 0
my_tumor = diz_appoggio[list(diz_appoggio.keys())[0]][0]
color = diz_appoggio[list(diz_appoggio.keys())[0]][1]
area = cv2.contourArea(my_tumor)
circumference = cv2.arcLength(my_tumor, True)
if circumference == 0:
    return "no tumor", 0
circularity = 4 * np.pi * (area / (circumference * circumference))
percentuale_area_tumore = (area / area_contorno_esterno) * 100
if circularity >= 0.75:
    probability =
    get_probability_of_tumor(area, circularity, color, colore_cervello, area_contorno_esterno)
    return my_tumor, probability
else:
    return "no tumor", 0
```

Segmentation

In this part we analyze the image returned from phase 1 and the value returned from phase 2, the k-mean algorithm is applied and the "best" contour is found

Trustworthiness of the program



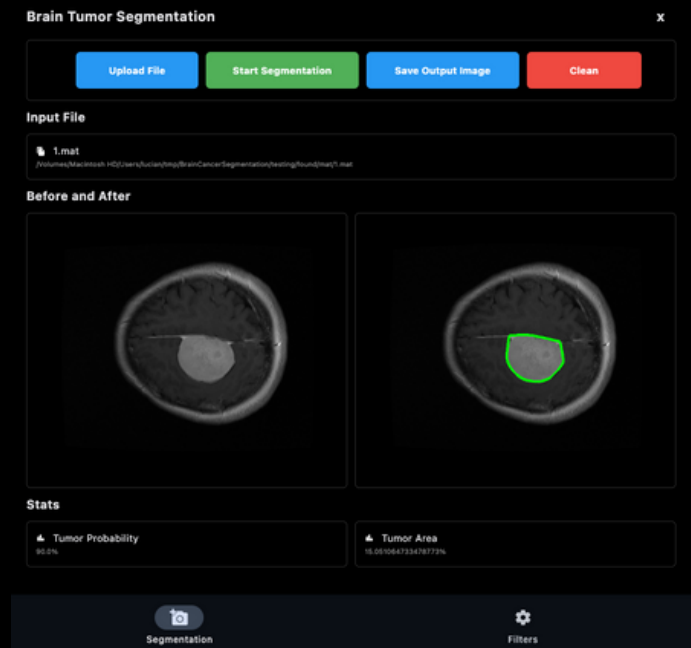
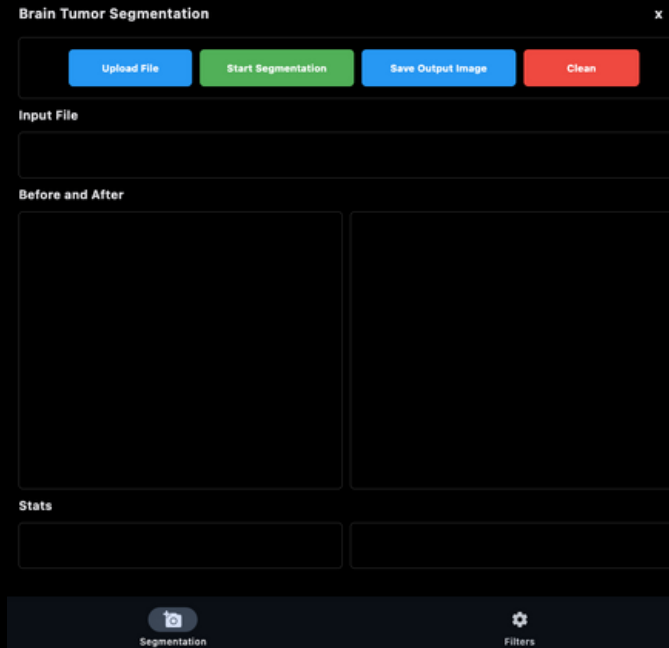
To get an estimate of how reliable the program was, tests were carried out by comparing the segment returned by the program with the "true" segment taken from the dataset.

Interface (GUI)

1. Brain Cancer Segmentation
2. Segmentation Filters

Brain Cancer Segmentation

The user can choose an MRI image of a patient's brain and begin segmentation



Segmentation Filters

The user can select at his choice different filters to apply to the image returned from the segmentation

Segmentation Filters

x

Upload Image

Save Output Image

Clean

Input File

23_35_23.png
/Volumes/Macintosh HD/Users/fucian/MyBrainCancerSegmentation/23_35_23.png

Select Options

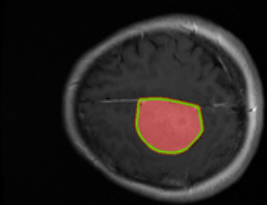
Increment/decrement light
+ 0 -

Color inside segmentation
Yes No
☒ ☐

Get only the tumor
Yes No
☐ ☒

Without segmentation
Yes No
☐ ☒

Generate Image



Success message
Congratulations, your image has been generated!

Segmentation

Filters

Segmentation Filters

x

Upload Image

Save Output Image

Clean

Input File

23_35_23.png
/Volumes/Macintosh HD/Users/fucian/MyBrainCancerSegmentation/23_35_23.png

Select Options


Increment/decrement light
+ 0 -

Color inside segmentation
Yes No
☐ ☒

Get only the tumor
Yes No
☒ ☐

Without segmentation
Yes No
☐ ☒

Generate Image



Success message
Congratulations, your image has been generated!

Segmentation

Filters



**Let's take a
look at
an example**