### In [24]:

```
import os
import warnings
warnings.simplefilter('ignore')
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
%matplotlib inline
from skimage.io import imread, imshow
from skimage.transform import resize
from skimage.color import rgb2gray
```

# In [25]:

```
ch=os.listdir("C:/Users/SRIVENKATESH/Downloads/CEC/chair")
ca=os.listdir("C:/Users/SRIVENKATESH/Downloads/CEC/car")
el=os.listdir("C:/Users/SRIVENKATESH/Downloads/CEC/elephant")
```

#### In [26]:

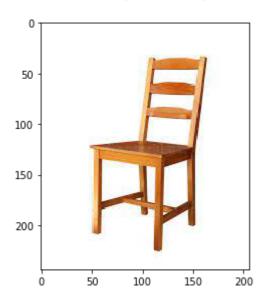
```
limit=10
ch_image=[None]*limit
j=0
for i in ch:
    if(j<limit):</pre>
        ch_image[j]=imread("C:/Users/SRIVENKATESH/Downloads/CEC/chair/"+i)
        j+=1
    else:
        break
limit=10
ca_image=[None]*limit
j=0
for i in ca:
    if(j<limit):</pre>
        ca_image[j]=imread("C:/Users/SRIVENKATESH/Downloads/CEC/car/"+i)
        j+=1
    else:
        break
    limit=10
el image=[None]*limit
j=0
for i in el:
    if(j<limit):</pre>
        el_image[j]=imread("C:/Users/SRIVENKATESH/Downloads/CEC/elephant/"+i)
        j+=1
    else:
        break
```

### In [27]:

imshow(ch\_image[4])

# Out[27]:

<matplotlib.image.AxesImage at 0x20ac32205b0>



# In [28]:

imshow(ca\_image[4])

### Out[28]:

<matplotlib.image.AxesImage at 0x20ac35d9670>

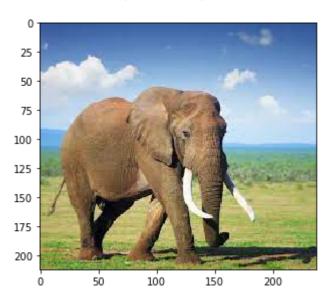


### In [29]:

```
imshow(el_image[4])
```

# Out[29]:

<matplotlib.image.AxesImage at 0x20abded20d0>



### In [30]:

ch\_image[4].shape

### Out[30]:

(244, 206, 3)

# In [31]:

ca\_image[4].shape

### Out[31]:

(168, 300, 3)

### In [32]:

el\_image[4].shape

### Out[32]:

(212, 238, 3)

#### In [33]:

```
ch_gray=[None]*limit
j=0
for i in ch:
    if(j<limit):</pre>
         ch_gray[j]=rgb2gray(ch_image[j])
         j+=1
    else:
         break
ca_gray=[None]*limit
j=0
for i in ca:
    if(j<limit):</pre>
         ca_gray[j]=rgb2gray(ca_image[j])
    else:
         break
el_gray=[None]*limit
j=0
for i in el:
    if(j<limit):</pre>
         el_gray[j]=rgb2gray(el_image[j])
         j+=1
    else:
         break
```

#### In [34]:

```
imshow(ch_gray[4])
```

#### Out[34]:

<matplotlib.image.AxesImage at 0x20abdf33cd0>



### In [35]:

imshow(ca\_gray[4])

# Out[35]:

<matplotlib.image.AxesImage at 0x20abe0f32b0>

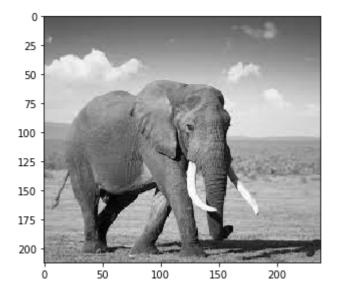


# In [36]:

imshow(el\_gray[4])

# Out[36]:

<matplotlib.image.AxesImage at 0x20ac01384f0>



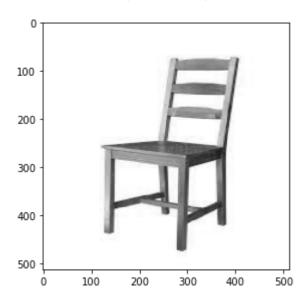
```
In [37]:
ch_gray[4].shape
Out[37]:
(244, 206)
In [38]:
ca_gray[4].shape
Out[38]:
(168, 300)
In [39]:
el_gray[4].shape
Out[39]:
(212, 238)
In [43]:
for j in range(10):
    ch=ch_gray[j]
    ch_gray[j]=resize(ch,(512,512))
In [44]:
for j in range(10):
    ca=ca_gray[j]
    ca_gray[j]=resize(ca,(512,512))
In [45]:
for j in range(10):
    el=el_gray[j]
    el_gray[j]=resize(el,(512,512))
```

# In [46]:

imshow(ch\_gray[4])

# Out[46]:

<matplotlib.image.AxesImage at 0x20ac01a6130>

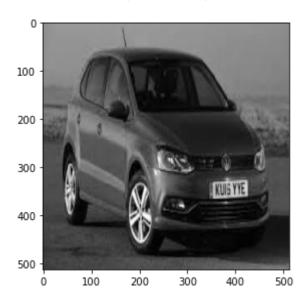


# In [47]:

imshow(ca\_gray[4])

# Out[47]:

<matplotlib.image.AxesImage at 0x20abdf9a160>

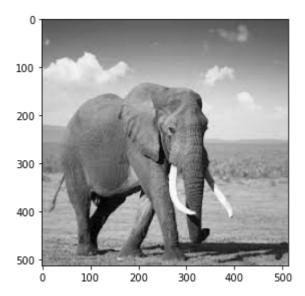


# In [48]:

imshow(el\_gray[4])

### Out[48]:

<matplotlib.image.AxesImage at 0x20abdffd130>



```
In [49]:
```

```
len_of_image_ch=len(ch_gray)
len_of_image_ch
len_of_image_ca=len(ca_gray)
len_of_image_ca
len_of_image_el=len(el_gray)
len_of_image_el
```

### Out[49]:

10

#### In [50]:

```
image_size_ca=ca_gray[4].shape
image_size_ca
image_size_el=el_gray[4].shape
image_size_el
image_size_ch=ch_gray[4].shape
image_size_ch=ch_gray[4].shape
image_size_ch
```

#### Out[50]:

(512, 512)

#### In [51]:

```
flatten_size_ca=image_size_ca[0]*image_size_ca[1]
flatten_size_ca
flatten_size_el=image_size_el[0]*image_size_el[1]
flatten_size_el
flatten_size_ch=image_size_ch[0]*image_size_ch[1]
flatten_size_ch
```

#### Out[51]:

262144

#### In [52]:

```
for i in range(len_of_image_ca):
    ca_gray[i]=np.ndarray.flatten(ca_gray[i]).reshape(flatten_size_ca,1)
ca_gray[4].shape
for i in range(len_of_image_el):
    el_gray[i]=np.ndarray.flatten(el_gray[i]).reshape(flatten_size_el,1)
el_gray[4].shape
for i in range(len_of_image_ch):
    ch_gray[i]=np.ndarray.flatten(ch_gray[i]).reshape(flatten_size_ch,1)
ch_gray[4].shape
```

#### Out[52]:

(262144, 1)

#### In [53]:

```
ca_gray=np.dstack(ca_gray)
el_gray=np.dstack(el_gray)
ch_gray=np.dstack(ch_gray)
```

#### In [54]:

```
ca_gray=np.rollaxis(ca_gray,axis=2,start=0)
ca_gray.shape
el_gray=np.rollaxis(el_gray,axis=2,start=0)
el_gray.shape
ch_gray=np.rollaxis(ch_gray,axis=2,start=0)
ch_gray.shape
```

#### Out[54]:

(10, 262144, 1)

#### In [56]:

```
ca_gray=ca_gray.reshape(len_of_image_ca,flatten_size_ca)
ca_gray.shape
el_gray=el_gray.reshape(len_of_image_el,flatten_size_el)
el_gray.shape
ch_gray=ch_gray.reshape(len_of_image_ch,flatten_size_ch)
ch_gray.shape
```

#### Out[56]:

(10, 262144)

# In [57]:

```
ca_data=pd.DataFrame(ca_gray)
ca_data
```

# Out[57]:

	0	1	2	3	4	5	6	7	8	
0	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
1	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	С
2	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	(
3	0.943523	0.944489	0.951485	0.946030	0.935423	0.933267	0.932656	0.938787	0.941648	C
4	0.561135	0.561135	0.561135	0.563294	0.565056	0.565056	0.566266	0.568564	0.570862	С
5	0.630525	0.632752	0.638669	0.646816	0.652649	0.655925	0.654333	0.650120	0.640460	C
6	0.225986	0.245332	0.306594	0.304438	0.287543	0.238678	0.209466	0.193071	0.185980	С
7	0.432715	0.443749	0.374935	0.339013	0.396952	0.445157	0.404925	0.364693	0.381725	C
8	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
9	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	С

### 10 rows × 262144 columns

**→** 

# In [58]:

el\_data=pd.DataFrame(el\_gray)
el\_data

### Out[58]:

	0	1	2	3	4	5	6	7	8	
0	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
1	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
2	0.911239	0.910965	0.908327	0.909763	0.913341	0.920210	0.928124	0.935630	0.943071	С
3	0.762583	0.762791	0.764226	0.764900	0.765274	0.767148	0.769255	0.770306	0.771415	С
4	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	С
5	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	(
6	0.968123	0.967495	0.968642	0.969790	0.970424	0.970817	0.971157	0.971157	0.971157	С
7	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	С
8	0.996451	0.994424	0.985269	0.980963	0.982586	0.985624	0.984999	0.979978	0.981284	С
9	0.421518	0.423566	0.418915	0.398364	0.373263	0.360707	0.372740	0.397506	0.399863	С

### 10 rows × 262144 columns

**→** 

# In [59]:

```
ch_data=pd.DataFrame(ch_gray)
ch_data
```

# Out[59]:

	0	1	2	3	4	5	6	7	8	
0	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
1	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
2	0.659031	0.659031	0.659031	0.659170	0.660769	0.662368	0.662670	0.662670	0.663593	С
3	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
4	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
5	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	С
6	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
7	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
8	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
9	0.604165	0.603629	0.605322	0.606972	0.606972	0.606972	0.608171	0.609864	0.611556	С

### 10 rows × 262144 columns

**→** 

# In [60]:

```
ca_data["label"]="ca"
ca_data
```

### Out[60]:

	0	1	2	3	4	5	6	7	8	
0	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
1	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	С
2	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	(
3	0.943523	0.944489	0.951485	0.946030	0.935423	0.933267	0.932656	0.938787	0.941648	С
4	0.561135	0.561135	0.561135	0.563294	0.565056	0.565056	0.566266	0.568564	0.570862	С
5	0.630525	0.632752	0.638669	0.646816	0.652649	0.655925	0.654333	0.650120	0.640460	С
6	0.225986	0.245332	0.306594	0.304438	0.287543	0.238678	0.209466	0.193071	0.185980	С
7	0.432715	0.443749	0.374935	0.339013	0.396952	0.445157	0.404925	0.364693	0.381725	С
8	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
9	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	С

### 10 rows × 262145 columns

**→** 

# In [61]:

```
el_data["label"]="el"
el_data
```

# Out[61]:

	0	1	2	3	4	5	6	7	8	
0	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
1	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
2	0.911239	0.910965	0.908327	0.909763	0.913341	0.920210	0.928124	0.935630	0.943071	С
3	0.762583	0.762791	0.764226	0.764900	0.765274	0.767148	0.769255	0.770306	0.771415	С
4	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	С
5	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	(
6	0.968123	0.967495	0.968642	0.969790	0.970424	0.970817	0.971157	0.971157	0.971157	C
7	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	C
8	0.996451	0.994424	0.985269	0.980963	0.982586	0.985624	0.984999	0.979978	0.981284	C
9	0.421518	0.423566	0.418915	0.398364	0.373263	0.360707	0.372740	0.397506	0.399863	С

# 10 rows × 262145 columns

**→** 

# In [62]:

```
ch_data["label"]="ch"
ch_data
```

### Out[62]:

	0	1	2	3	4	5	6	7	8	
0	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
1	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
2	0.659031	0.659031	0.659031	0.659170	0.660769	0.662368	0.662670	0.662670	0.663593	С
3	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
4	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
5	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	С
6	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
7	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
8	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
9	0.604165	0.603629	0.605322	0.606972	0.606972	0.606972	0.608171	0.609864	0.611556	С

### 10 rows × 262145 columns

**→** 

# In [63]:

```
actor_1=pd.concat([ca_data,el_data,ch_data])
actor_1
```

# Out[63]:

	0	1	2	3	4	5	6	7	8	
0	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
1	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	С
2	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	(
3	0.943523	0.944489	0.951485	0.946030	0.935423	0.933267	0.932656	0.938787	0.941648	С
4	0.561135	0.561135	0.561135	0.563294	0.565056	0.565056	0.566266	0.568564	0.570862	С
5	0.630525	0.632752	0.638669	0.646816	0.652649	0.655925	0.654333	0.650120	0.640460	С
6	0.225986	0.245332	0.306594	0.304438	0.287543	0.238678	0.209466	0.193071	0.185980	С
7	0.432715	0.443749	0.374935	0.339013	0.396952	0.445157	0.404925	0.364693	0.381725	С
8	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
9	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	С
0	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
1	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
2	0.911239	0.910965	0.908327	0.909763	0.913341	0.920210	0.928124	0.935630	0.943071	С
3	0.762583	0.762791	0.764226	0.764900	0.765274	0.767148	0.769255	0.770306	0.771415	С
4	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	С
5	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	C
6	0.968123	0.967495	0.968642	0.969790	0.970424	0.970817	0.971157	0.971157	0.971157	С
7	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	С
8	0.996451	0.994424	0.985269	0.980963	0.982586	0.985624	0.984999	0.979978	0.981284	С
9	0.421518	0.423566	0.418915	0.398364	0.373263	0.360707	0.372740	0.397506	0.399863	С
0	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
1	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
2	0.659031	0.659031	0.659031	0.659170	0.660769	0.662368	0.662670	0.662670	0.663593	С
3	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
4	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
5	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	С
6	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
7	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
8	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1
9	0.604165	0.603629	0.605322	0.606972	0.606972	0.606972	0.608171	0.609864	0.611556	С

30 rows × 262145 columns

# In [64]:

```
from sklearn.utils import shuffle
celebrities_indexed=shuffle(actor_1).reset_index()
celebrities_indexed
```

# Out[64]:

	index	0	1	2	3	4	5	6	7	
0	7	0.432715	0.443749	0.374935	0.339013	0.396952	0.445157	0.404925	0.364693	0.3
1	1	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.0
2	4	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.0
3	9	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.7
4	5	0.630525	0.632752	0.638669	0.646816	0.652649	0.655925	0.654333	0.650120	0.6
5	9	0.421518	0.423566	0.418915	0.398364	0.373263	0.360707	0.372740	0.397506	0.3
6	7	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.0
7	6	0.968123	0.967495	0.968642	0.969790	0.970424	0.970817	0.971157	0.971157	0.9
8	0	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.0
9	8	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.0
10	3	0.943523	0.944489	0.951485	0.946030	0.935423	0.933267	0.932656	0.938787	0.9
11	9	0.604165	0.603629	0.605322	0.606972	0.606972	0.606972	0.608171	0.609864	0.6
12	2	0.659031	0.659031	0.659031	0.659170	0.660769	0.662368	0.662670	0.662670	0.6
13	4	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.3
14	1	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.9
15	8	0.996451	0.994424	0.985269	0.980963	0.982586	0.985624	0.984999	0.979978	0.9
16	8	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.0
17	2	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.1
18	0	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.0
19	5	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.9
20	2	0.911239	0.910965	0.908327	0.909763	0.913341	0.920210	0.928124	0.935630	0.9
21	6	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.0
22	0	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.0
23	5	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	8.0
24	7	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.5
25	6	0.225986	0.245332	0.306594	0.304438	0.287543	0.238678	0.209466	0.193071	0.1
26	3	0.762583	0.762791	0.764226	0.764900	0.765274	0.767148	0.769255	0.770306	0.7
27	1	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.0
28	4	0.561135	0.561135	0.561135	0.563294	0.565056	0.565056	0.566266	0.568564	0.5
29	3	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.0

30 rows × 262146 columns



### In [65]:

```
celebrities_actor_1=celebrities_indexed.drop(['index'],axis=1)
celebrities_actor_1
```

# Out[65]:

	0	1	2	3	4	5	6	7	8
0	0.432715	0.443749	0.374935	0.339013	0.396952	0.445157	0.404925	0.364693	0.381725
1	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
2	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
3	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658	0.763658
4	0.630525	0.632752	0.638669	0.646816	0.652649	0.655925	0.654333	0.650120	0.640460
5	0.421518	0.423566	0.418915	0.398364	0.373263	0.360707	0.372740	0.397506	0.399863
6	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
7	0.968123	0.967495	0.968642	0.969790	0.970424	0.970817	0.971157	0.971157	0.971157
8	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
9	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
10	0.943523	0.944489	0.951485	0.946030	0.935423	0.933267	0.932656	0.938787	0.941648
11	0.604165	0.603629	0.605322	0.606972	0.606972	0.606972	0.608171	0.609864	0.611556
12	0.659031	0.659031	0.659031	0.659170	0.660769	0.662368	0.662670	0.662670	0.663593
13	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931	0.339931
14	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550	0.988550
15	0.996451	0.994424	0.985269	0.980963	0.982586	0.985624	0.984999	0.979978	0.981284
16	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
17	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776	0.111776
18	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
19	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020	0.949020
20	0.911239	0.910965	0.908327	0.909763	0.913341	0.920210	0.928124	0.935630	0.943071
21	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
22	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
23	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747	0.811747
24	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873	0.540873
25	0.225986	0.245332	0.306594	0.304438	0.287543	0.238678	0.209466	0.193071	0.185980
26	0.762583	0.762791	0.764226	0.764900	0.765274	0.767148	0.769255	0.770306	0.771415
27	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000
28	0.561135	0.561135	0.561135	0.563294	0.565056	0.565056	0.566266	0.568564	0.570862
29	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000	1.000000

30 rows × 262145 columns

```
In [66]:
celebrities_actor_1.to_csv("C:/Users/SRIVENKATESH/Downloads/CEC.csv")
In [67]:
x=celebrities_actor_1.values[:,:-1]
y=celebrities_actor_1.values[:,-1]
In [68]:
х
Out[68]:
array([[0.4327151439898622, 0.4437485868723252, 0.3749347457369636, ...,
       0.5621746139705883, 0.5639898713235294, 0.5636988174019608],
       [1.0, 1.0, 1.0, \ldots, 1.0, 1.0, 1.0],
       [1.0, 1.0, 1.0, \ldots, 1.0, 1.0, 1.0],
       [1.0, 1.0, 1.0, \ldots, 1.0, 1.0, 1.0],
       [0.5611345098039217, 0.5611345098039217, 0.5611345098039217, ...,
       0.46630876531862747, 0.46630876531862747, 0.4663087653186274],
       [1.0, 1.0, 1.0, ..., 1.0, 1.0, 1.0]], dtype=object)
In [69]:
У
Out[69]:
In [75]:
from sklearn.model selection import train test split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=0)
In [76]:
from sklearn import svm
In [77]:
clf=svm.SVC()
clf.fit(x_train, y_train)
Out[77]:
```

SVC()

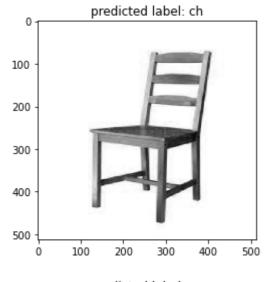
#### In [78]:

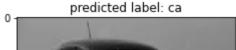
```
y_pred=clf.predict(x_test)
y_pred
```

#### Out[78]:

#### In [80]:

```
for i in range(10):
    predicted_images=(np.reshape(x_test[i],(512,512)).astype(np.float64))
    plt.title('predicted label: {0}'. format(y_pred[i]))
    plt.imshow(predicted_images, interpolation='nearest', cmap='gray')
    plt.show()
```





#### In [81]:

```
from sklearn import metrics
accuracy=metrics.accuracy_score(y_test,y_pred)
accuracy
```

#### Out[81]:

#### 0.466666666666667

#### In [82]:

```
from sklearn.metrics import confusion_matrix
confusion_matrix(y_test,y_pred)
```

#### Out[82]:

In [ ]:		