

## Let's use deep learning to find out similar images.

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we will use K nearest neighbour algorithm.

```
In [1]: import graphlab

In [2]: # Limit number of worker processes. This preserves system memory, which prevents hosted notebooks from crashing.
graphlab.set_runtime_config('GRAPHLAB_DEFAULT_NUM_PYLAMBDA_WORKERS', 4)

[INFO] graphlab.cython.cy_server: GraphLab Create v2.1 started. Logging: /tmp/graphlab_server_1481433924.log

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In [9]: # load the data for images.

image_train = graphlab.SFrame('image_train_data/')

In [5]: # view data.
graphlab.canvas.set_target('ipynb')
image_train.show()
```

id		image		label		deep_features		image_array			
dtype:	int	dtype:	Image	dtype:	str	dtype:	array	dtype:	array		
num_unique (est.):	1,999	First 4 images: 		num_unique (est.):	4	num_unique (est.):	2,336,740	num_unique (est.):	255		
num_undefined:	0			num_undefined:	0	num_undefined:	0	num_undefined:	0		
min:	24			frequent items: <table><tr><td>automobile</td></tr><tr><td>dog</td></tr><tr><td>cat</td></tr><tr><td>bird</td></tr></table>		automobile	dog	cat	bird	min:	0
automobile											
dog											
cat											
bird											
max:	49,970			max:	15.345	max:	15.345	max:	255		
median:	23,969			median:	0	median:	0	median:	113		
mean:	24,828.162			mean:	0.386	mean:	0.386	mean:	116.98		
std:	14,682.469			std:	0.905	std:	0.905	std:	64.312		
distribution of values: 						distribution of values (all sub-columns): 		distribution of values (all sub-columns): 			

```
In [8]: # train the model using nearest-neighbour algorithm.

knn_model = graphlab.nearest_neighbors.create(image_train,
                                              features=['deep_features'],
                                              label='id')

Starting brute force nearest neighbors model training.

In [18]: # now let's pick 4th image as its cat as we show in above show() command.

test1=image_train[3:4]
test1['image'].show()
```

All 1 images in <SArray>



```
In [19]: # now let's find out similar images.
knn_model.query(test1)

Starting pairwise querying.

+-----+-----+-----+-----+
| Query points | # Pairs | % Complete. | Elapsed Time |
+-----+-----+-----+-----+
```

```
| 0          | 1          | 0.0498753 | 19.527ms |
| Done       |            | 100        | 251.811ms |
```

Out[19]:

query_label	reference_label	distance	rank
0	70	0.0	1
0	19437	37.563553912	2
0	41989	40.4601125879	3
0	40249	40.9408480727	4
0	11000	41.3051067894	5

[5 rows x 4 columns]

In [20]: *# we don't see the images, so that's not easy to understand.*





```
res = knn_model.query(test1)
```

Starting pairwise querying.

```
+-----+-----+-----+-----+
| Query points | # Pairs | % Complete. | Elapsed Time |
+-----+-----+-----+-----+
| 0          | 1          | 0.0498753 | 33.866ms |
| Done       |            | 100        | 257.816ms |
+-----+-----+-----+-----+
```

In [21]: *# now let's retrieve the image.*

```
image_train.filter_by(res['reference_label'], 'id').show()
```

id		image		label		deep_features		image_array	
dtype:	int	dtype:	Image	dtype:	str	dtype:	array	dtype:	array
num_unique (est.):	5	First 4 images: 		num_unique (est.):	3	num_unique (est.):	6,762	num_unique (est.):	249
num_undefined:	0			num_undefined:	0	num_undefined:	0	num_undefined:	0
min:	70			frequent items: <div>dog</div> <div>cat</div> <div>automobile</div>		min:	0	min:	0
max:	41,989					max:	11.237	max:	255
median:	19,437					median:	0	median:	109
mean:	22,549					mean:	0.424	mean:	108.864
std:	16,368.097					std:	0.933	std:	61.72
distribution of values:						distribution of values (all sub-columns):		distribution of values (all sub-columns):	
									

In [25]: *# good results, but not perfect.*

*# let's create a function to make it easier to find neighbours and find out the quality of results.*

```
def show_neighbours(i):
    res = knn_model.query(image_train[i:i+1])
    fil_res = image_train.filter_by(res['reference_label'], 'id')
    fil_res['image'].show()
```

In [26]: show\_neighbours(13)

Starting pairwise querying.

```
+-----+-----+-----+-----+
| Query points | # Pairs | % Complete. | Elapsed Time |
+-----+-----+-----+-----+
| 0          | 1          | 0.0498753 | 24.224ms |
| Done       |            | 100        | 218.68ms |
+-----+-----+-----+-----+
```

All 5 images in <SArray>



In [ ]: *# nice. First image is the query and it is matching with all cars with almost similar white color.*

```
# let's play bit more.
```

```
In [29]: show_neighbours(1537)
```

Starting pairwise querying.

Query points	# Pairs	% Complete.	Elapsed Time
0	1	0.0498753	25.668ms
Done		100	233.316ms

All 5 images in <SArray>



```
In [ ]: # for car, the results are coming out good. For rest it has good success.
```