## Machine Learning Assignment 3

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### Looking at metadata for features:

)	pl	given location on the planar dimensions
Į.	si	variations in height, width, area
2	va	the various degrees between white and black
3	te	variation in the fineness or coarseness of an area having a given value; includes blur
Ē	со	hue, using the repertoire of colored sensations which can be produced at equal value
j	or	various orientations, ranging from the vertical to the horizontal in a distinct direction
j	sh	a mark with a constant size can nonetheless have an infinite number of different shapes
7	reflection	indicates the work contains an image given back by a reflecting surface, or an image seen in a mirror or shiny surface
3	ро	A POINT represents a location on the plane that has no theoretical length or area. This signification is independent of the size and character of the mai
)	li	A LINE signifies a phenomenon on the plane which has measurable length but no area. This signification is independent of the width and characteristic
3	ar	An AREA signifies something on the plane that has a measurable size. This signification applies to the entire area covered by the visible mark.
	notes	notes, description

In order to read in the metadata, we will need to transform the datatype.

#### Out[30]:

notes	reflection	date	country_of_origin	artist	
distorted perspective, shadow, signification o	False distorted perspective, shadow, signification	1913	Giorgio de Chirico Italy		0
hard to understand the viewpoint, sense of for	False	1967	Italy	Giovanni Anselmo	1
flatish, textured shapes & specific colors-lin	False	1958	America	Milton Avery	2
shapes, layers, paint handlng/texture, orienta	False	1957	UK	Gillian Avery	3
ambiguity through abstraction, odd shape, v te	False	1956	France	Joseph (Jef) Banc	4

ValueError: could not convert string to float: 'shapes placed separately on ground, vary in s ize, sense of liquid paint, shapes & lines bleed / merge, ambiguity through abstraction, shap es are somewhat moorless & float in relation to each other, what is the space, but less about uncertainty'

Part 1 to change data type of reflection from Bool to int64 using astype()

```
# Transform the data type values of 'reflection' from bool True/False to to int64 '0', or '1' F
# Find the data type of Reflection Column
reflection_isbool = type(data.reflection[0])
# Convert to'int64'
data.reflection = data.reflection.astype('int64')
# Find data type after casting
after = type(data.reflection[0])
# Print new data type of 'reflection'
after
```

Part 2 Set (True, False) values to (1, 0)

Transform Data type for columns 'reflection' & 'has\_text'

#### Convert data type in Featured Columns

```
In [147]: # Convert the "reflection" and "has_text" colums from True/False to 0,1
    data['reflection'] = data['reflection'].astype(int)
    data['has_text'] = data['has_text'].astype(int)
```

has_text	pri	reflection
0		0
0		0
0		0
0		0
0		0
1		0
0		0
0		0
0		0
0		1

### plot inertia scores by number of clusters

```
In [289]: # first attempt at fitting K means to view change in Inertia
           # class sklearn.cluster.KMeans(n_clusters=8, init='k-means++', n_init=10,
           # container to store inertia scores over iterations
           distortions = []
           # fit KMeans iteratively to begin to assess the appropriate number of clu
           for i in range(15, 19):
               km = KMeans(n_clusters=i)
               km.fit(X)
               distortions.append(km.inertia)
           # vizualize change in inertia
          plt.plot(range(15, 19), distortions, marker='o')
plt.xlabel('Number of clusters')
           plt.ylabel('Inertia')
           plt.show()
              1520
              1500
              1480
              1460
              1440
              1420
              1400
                                       16.5
                                                    17.5
                                                           18.0
                  15.0
                         15.5
                                16.0
                                             17.0
                                  Number of clusters
```

More Work is to be done on printing images, using IPython.display

Many of the images are ending up in multiple clusters, so my next iteration will incorporate additional changes, where each artwork can only be clustered in one place, also it will be critical to see the data (images) and search for patterns using different features.

### Cluster 0 Samples:

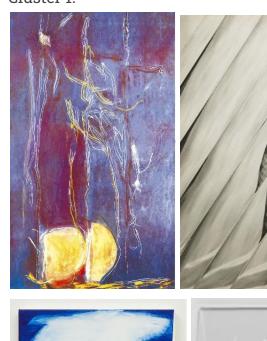








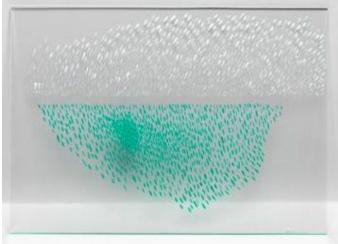
# Cluster 1:











## Cluster 2:





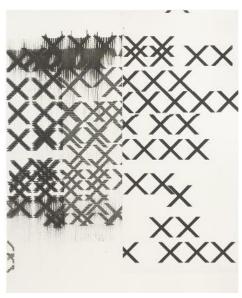


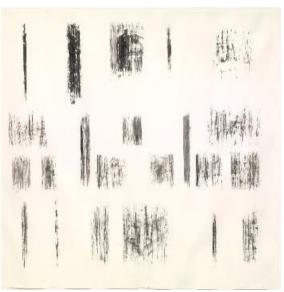
Cluster 3:

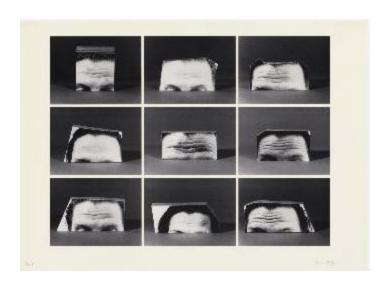












## Cluster 4:





