

Tarea2

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1 Proyecto 2: Propedéutico

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1.1 Parte 1: Teoría de Álgebra Lineal y Optimización

Necesito estudiar esta parte

1.2 Parte 2: Aplicaciones en Python

```
In [1]: import numpy as np
        from PIL import Image
        import matplotlib.pyplot #import matplotlib.pyplot as plt

        img = Image.open("tarea2.color.png")
        img
```

Out[1]:



```
In [2]: img = img.convert("LA")
        row, col = img.size
        data = np.zeros((row*col, 5))
        matplotlib.pyplot.figure(figsize=(9, 6))
        matplotlib.pyplot.imshow(img)
        img
```

Out [2]:



1.2.1 1. SVD

```
In [3]: A = np.matrix(img.getdata(band=0), float)
        A
```

```
Out[3]: matrix([[ 171.,  180.,  176., ...,  64.,  65.,  63.]])
```

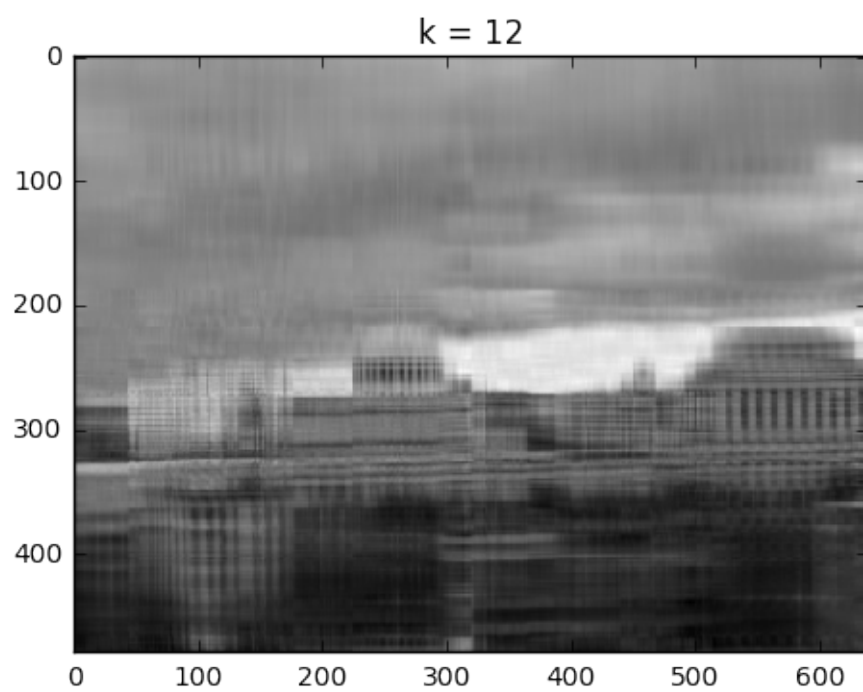
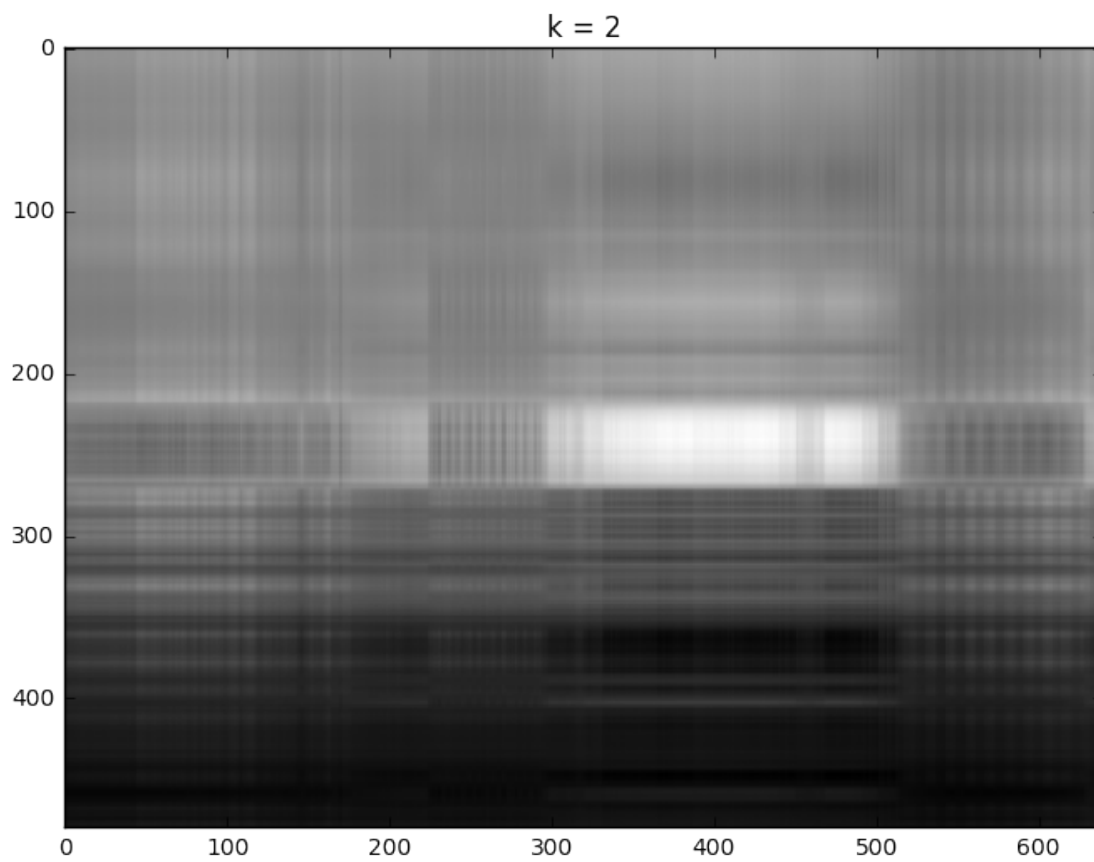
```
In [4]: A.shape = (img.size[1], img.size[0])
        A.size
```

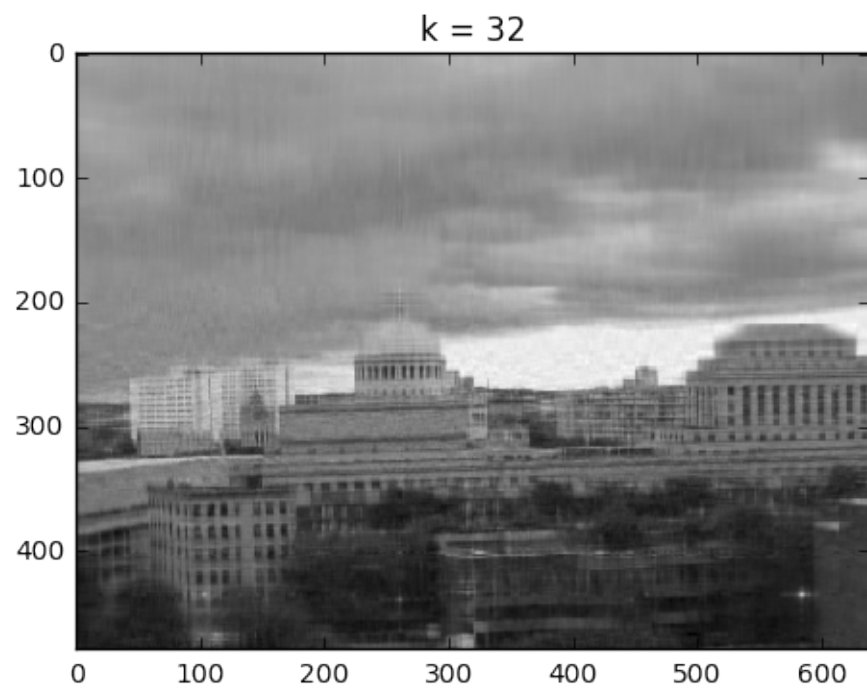
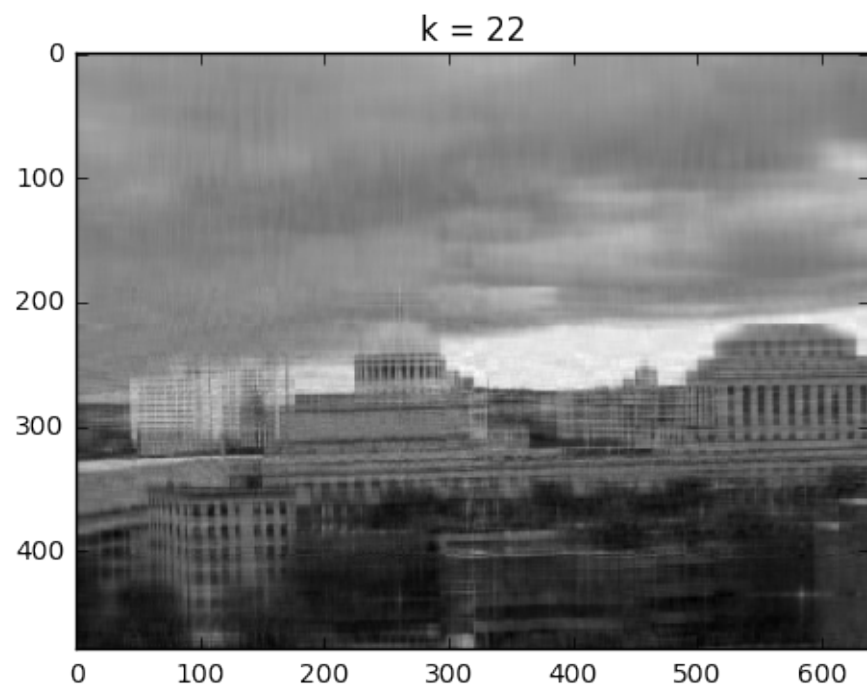
```
Out[4]: 307200
```

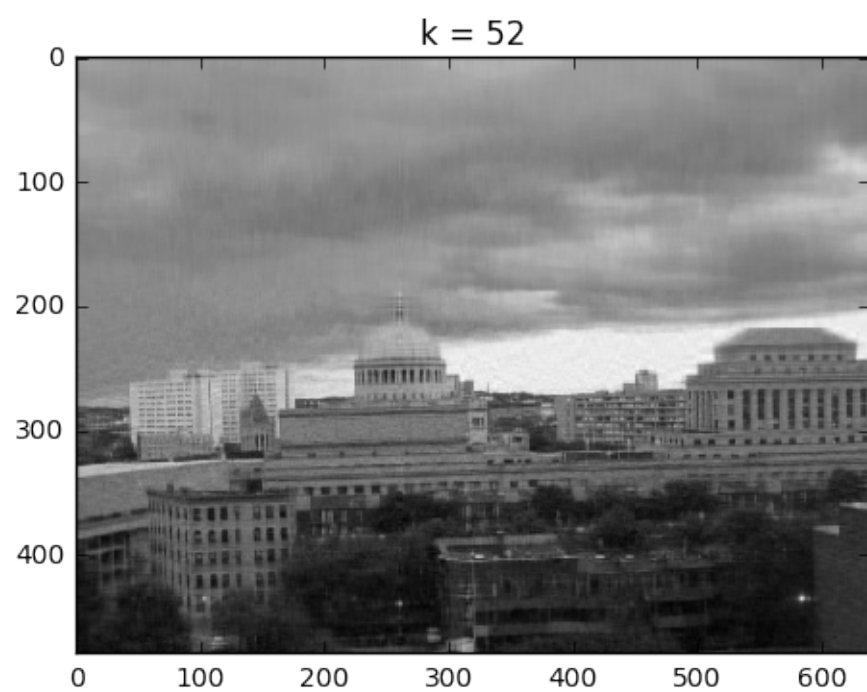
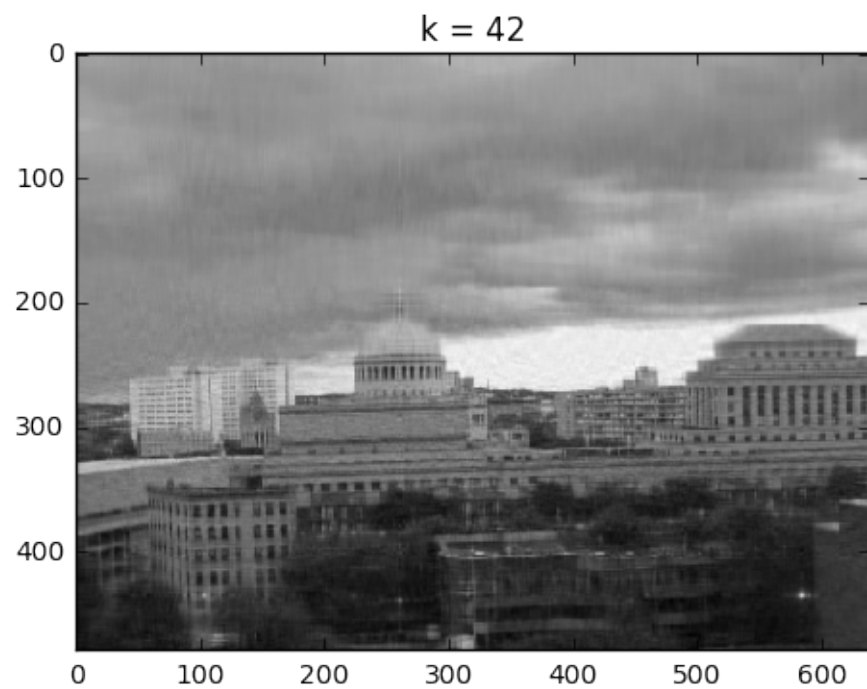
```
In [5]: u, s, v = np.linalg.svd(A, full_matrices=1)
```

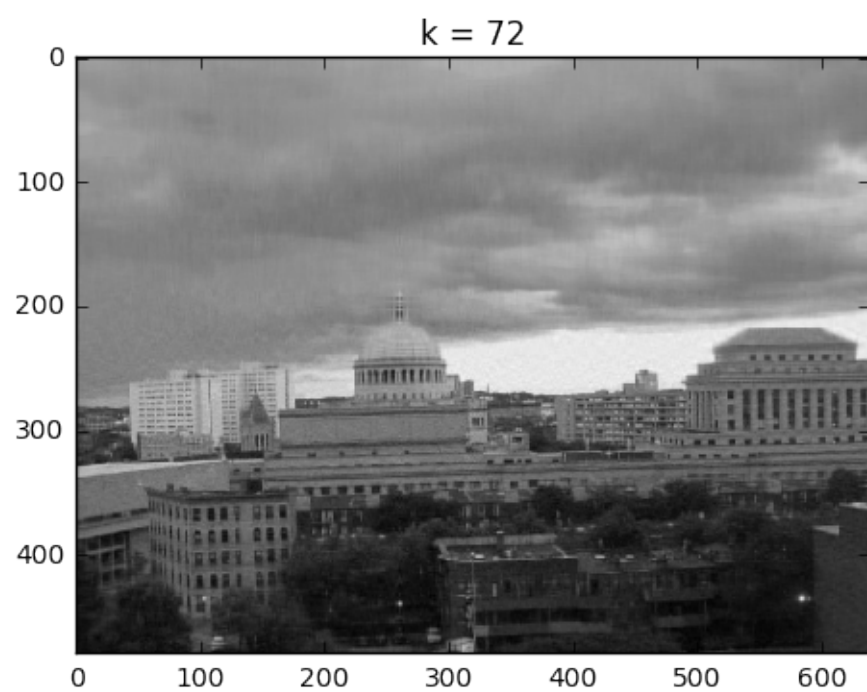
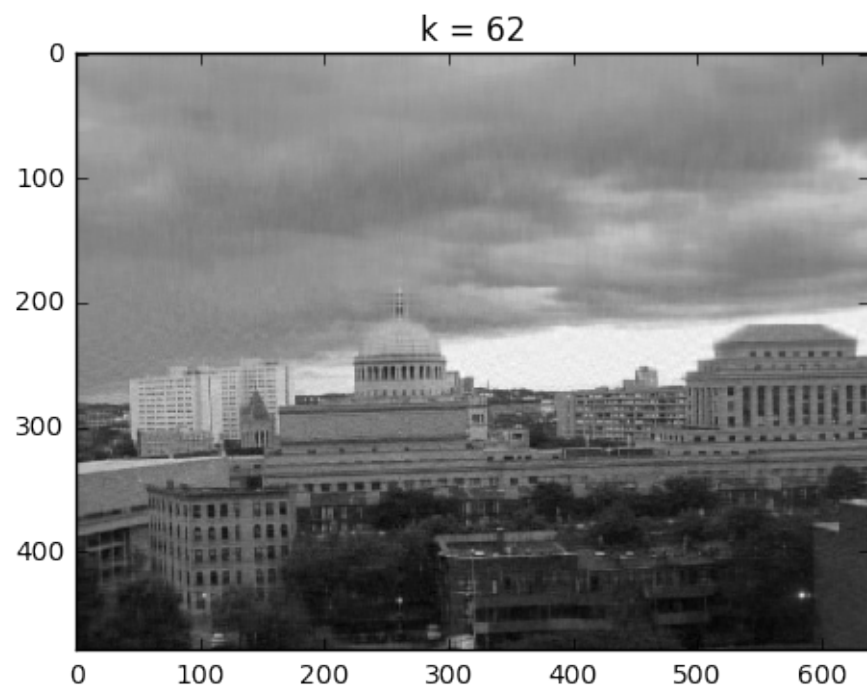
```
In [6]: matplotlib.pyplot.figure(figsize=(9,6))
        matplotlib.pyplot.imshow(img, cmap='gray')
        for k in range(2,100,10):
            B = np.matrix(u[:, :k]) * np.diag(s[:k]) * np.matrix(v[:,k, :])
            matplotlib.pyplot.imshow(B, cmap='gray')
            matplotlib.pyplot.title("k = " + str(k))
            matplotlib.pyplot.show()
```

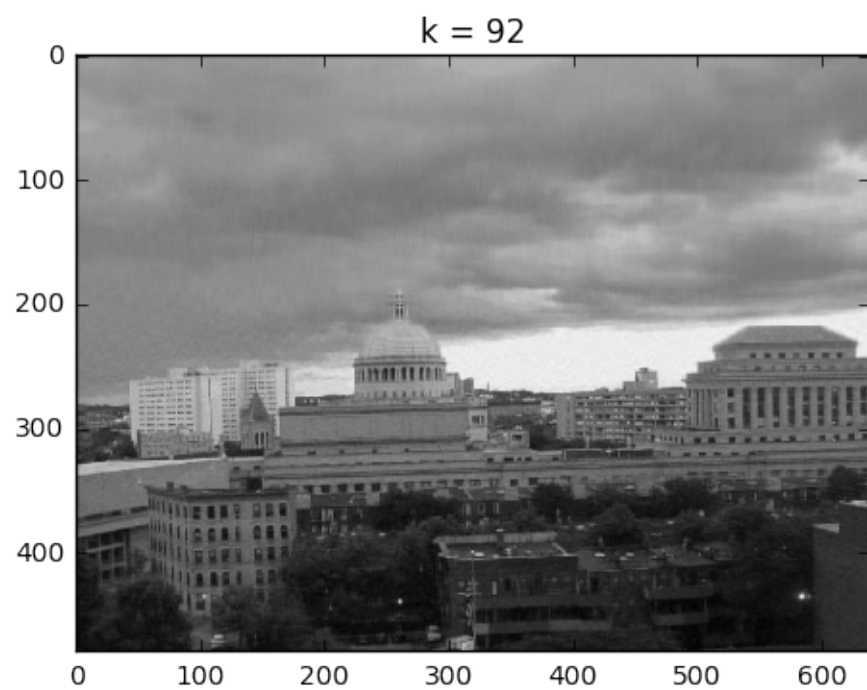
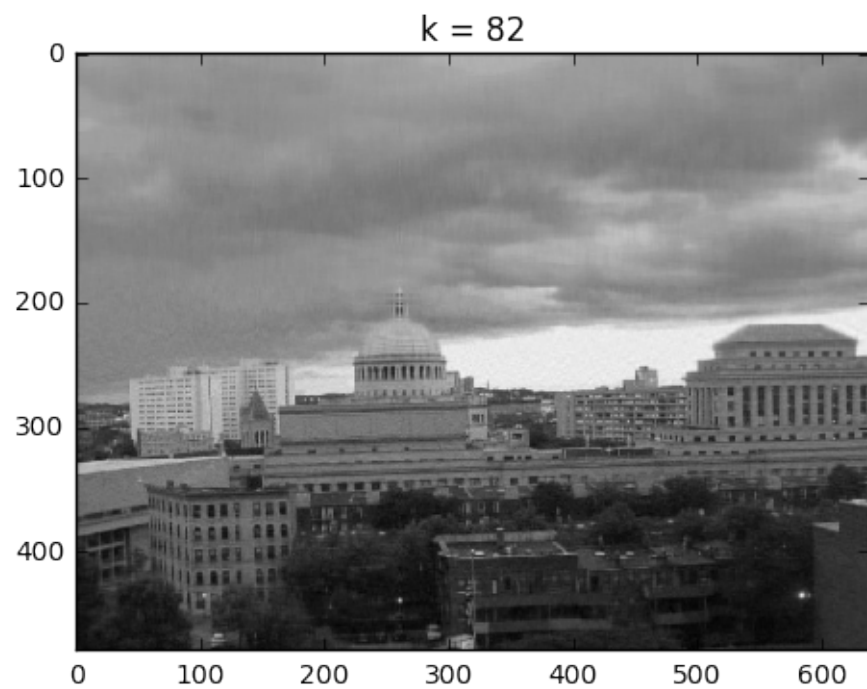












¿Qué tiene que ver con la compresión de imágenes? Como se puede observar, muchos algoritmos utilizan la compresión a través de la operación SVD, pues reduce el número de columnas, renglones en una matriz y va reduciéndose de manera progresiva el número de renglones y en consecuencia del tamaño de archivos, por ende perdiendo calidad.

1.2.2 a. Pseudoinversa

```
In [7]: import numpy as np

def pinversa(A):
    u, s, v = np.linalg.svd(A)
    B = np.dot(v.transpose(), np.dot(np.diag(s**-1), u.transpose()))
    B = v.transpose() * (np.diag(s**-1)*u.transpose())
    return np.array(B)

A = np.matrix([[1,2,3],[4,5,6],[7,8,9]])
C = np.matrix([[1,2],[4,5]])
C = pinversa(C)
C

Out[7]: array([[ -1.66666667,  0.66666667],
               [ 1.33333333, -0.33333333]])
```

Se muestra como al aplicar la inversa nuevamente vuelve a su forma original

```
In [8]: np.linalg.pinv(C)

Out[8]: array([[ 1.,  2.],
               [ 4.,  5.]])
```

1.2.3 3. Pandas

```
In [15]: import pandas as pd
import os
import numpy as np
import statsmodels.formula.api as sm

studysat = pd.read_csv("study_vs_sat.csv")
studysat = pd.DataFrame(studysat)
studysat

Out[15]:
```

	study_hours	sat_score
0	4	390
1	9	580
2	10	650
3	14	730
4	4	410
5	7	530
6	12	600

7	22	790
8	1	350
9	3	400
10	8	590
11	11	640
12	5	450
13	6	520
14	10	690
15	11	690
16	16	770
17	13	700
18	13	730
19	10	640

```
In [28]: modelo = sm.ols("study_hours ~ sat_score", studysat)
         modelo = modelo.fit()
         modelo.params
```

```
Out[28]: Intercept    -10.941095
         sat_score      0.034415
         dtype: float64
```

```
In [35]: sumhours = sum(studysat["study_hours"])
         sumsat = sum(studysat["sat_score"])
         sumhourssat = sum(studysat["study_hours"] * studysat["sat_score"])
         sumhours2 = sum(studysat["study_hours"]**2)
         sumsat2 = sum(studysat["sat_score"]**2)
         length = len(studysat["study_hours"])
         = (length * sumhourssat - sumhours*sumsat)/(length * sumhours2 - sumhours**2)
```

```
Out[35]: 25.326467777895743
```

```
In [36]: = sumsat / length - (*sumhours) / length
```

```
Out[36]: 353.16487949888523
```