

```

import pandas as pd
import numpy as np
from sklearn.metrics import mean_squared_error

alsi = pd.read_csv("Dataset/Mercury/alsimap_smooth_032015.png.csv",
header = None)
casi = pd.read_csv("Dataset/Mercury/casimap_smooth_032015.png.csv",
header = None)
fesi = pd.read_csv("Dataset/Mercury/fesimap_smooth_032015.png.csv",
header = None)
mgssi = pd.read_csv("Dataset/Mercury/mgsimap_smooth_032015.png.csv",
header = None)
ssi = pd.read_csv("Dataset/Mercury/ssimap_smooth_032015.png.csv",
header = None)
top = pd.read_csv("Dataset/Mercury/mercury-albedo-top-half.png.csv",
header = None)
bottom = pd.read_csv("Dataset/Mercury/mercury-albedo-resized-bottom-
half.png.csv", header = None)

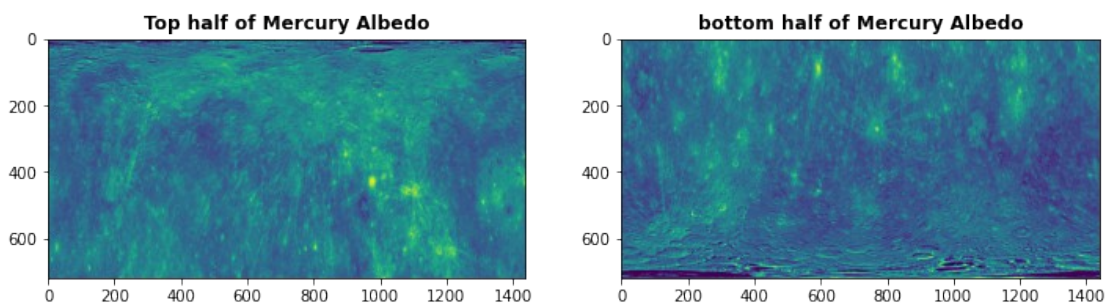
%matplotlib inline
import matplotlib.pyplot as plt

plt.figure(figsize =(12,12))
plt.subplot(1, 2, 1)
plt.imshow(top)
plt.title('Top half of Mercury Albedo', fontweight="bold")

plt.subplot(1, 2, 2)
plt.imshow(bottom)
plt.title('bottom half of Mercury Albedo', fontweight="bold")

Text(0.5, 1.0, 'bottom half of Mercury Albedo')

```



### Filling gaps

```

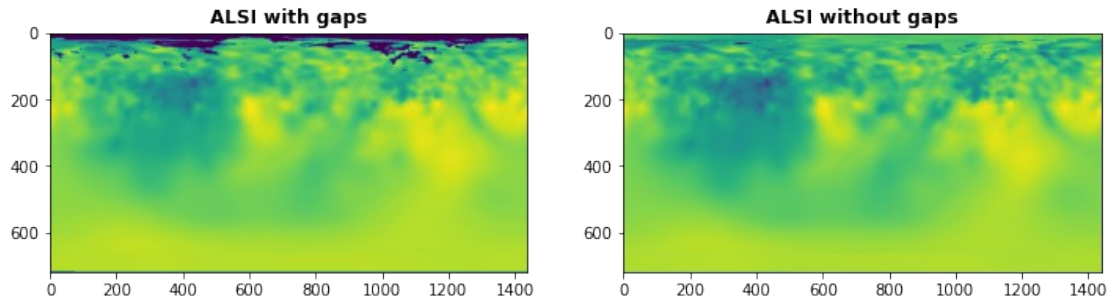
plt.figure(figsize =(12,12))
plt.subplot(1, 2, 1)
plt.imshow(alsi)
plt.title('ALSI with gaps', fontweight="bold")

alsi_mean = alsi.replace(0E+00, alsi.mean())

```

```
plt.subplot(1, 2, 2)
plt.imshow(alsi_mean)
plt.title('ALSI without gaps', fontweight="bold")

Text(0.5, 1.0, 'ALSI without gaps')
```



```
alsi_mean.std().sum()
```

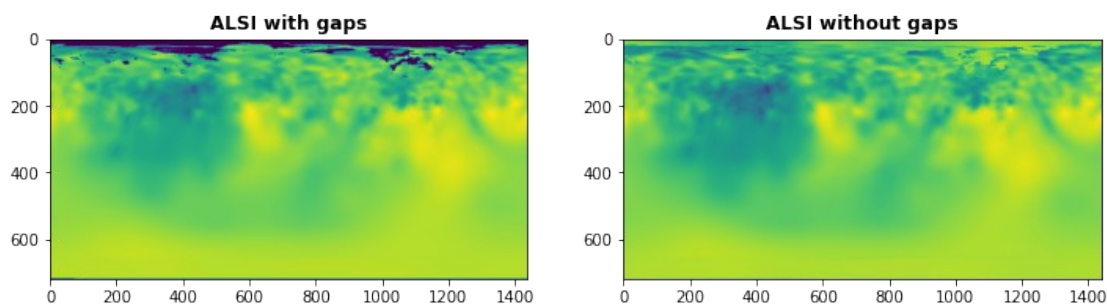
```
122.94801210708867
```

```
plt.figure(figsize=(12,12))
plt.subplot(1, 2, 1)
plt.imshow(alsi)
plt.title('ALSI with gaps', fontweight="bold")
```

```
alsi_median = alsi.replace(0E+00, alsi.median())
```

```
plt.subplot(1, 2, 2)
plt.imshow(alsi_median)
plt.title('ALSI without gaps', fontweight="bold")
```

```
Text(0.5, 1.0, 'ALSI without gaps')
```



```
alsi_median.std().sum()
```

```
122.29112638414719
```

On comparing the standard deviations of both data, imputing gaps with median is giving better results.

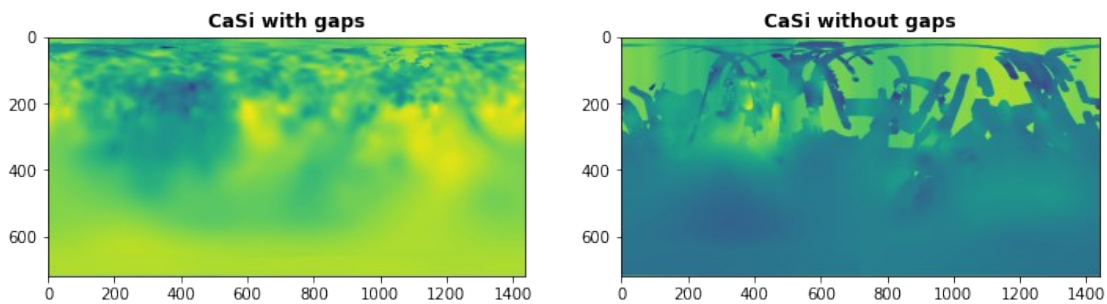
```
alsi = alsi_median
```

```
plt.figure(figsize =(12,12))
plt.subplot(1, 2, 1)
plt.imshow(alsi)
plt.title('CaSi with gaps', fontweight="bold")

casi = casi.replace(0E+00, alsi.median())

plt.subplot(1, 2, 2)
plt.imshow(casi)
plt.title('CaSi without gaps', fontweight="bold")

Text(0.5, 1.0, 'CaSi without gaps')
```

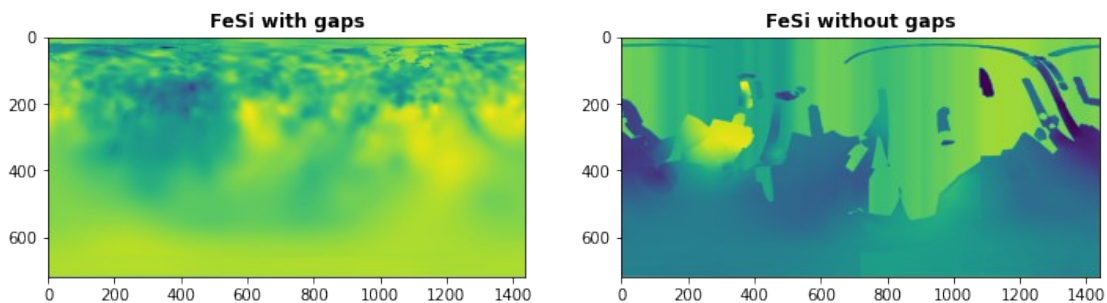


```
plt.figure(figsize =(12,12))
plt.subplot(1, 2, 1)
plt.imshow(alsi)
plt.title('FeSi with gaps', fontweight="bold")

fesi = fesi.replace(0E+00, alsi.median())

plt.subplot(1, 2, 2)
plt.imshow(fesi)
plt.title('FeSi without gaps', fontweight="bold")

Text(0.5, 1.0, 'FeSi without gaps')
```



```
plt.figure(figsize =(12,12))
plt.subplot(1, 2, 1)
plt.imshow(alsi)
plt.title('MgSi with gaps', fontweight="bold")
```

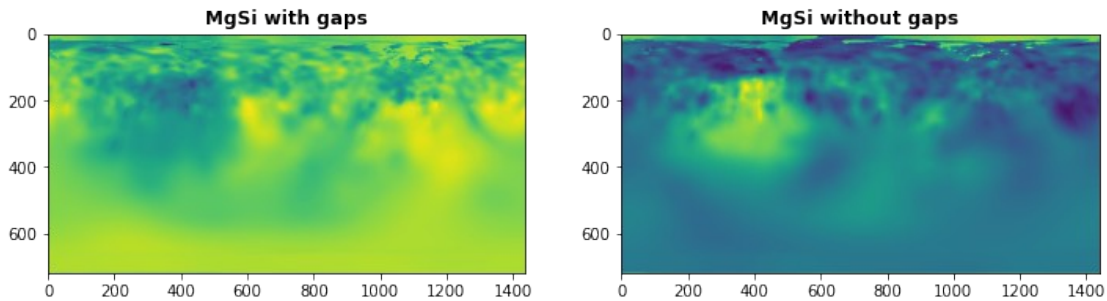
```

mgsi = mgsi.replace(0E+00, alsi.median())

plt.subplot(1, 2, 2)
plt.imshow(mgsi)
plt.title('MgSi without gaps', fontweight="bold")

Text(0.5, 1.0, 'MgSi without gaps')

```



```

plt.figure(figsize =(12,12))
plt.subplot(1, 2, 1)
plt.imshow(alsi)
plt.title('SSi with gaps', fontweight="bold")

```

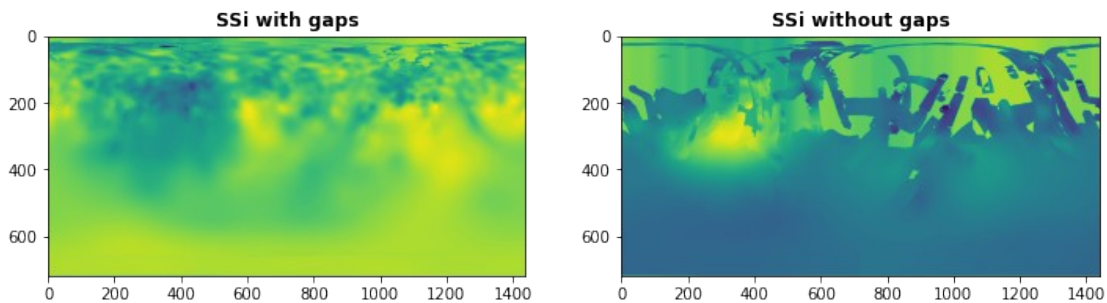
```

ssi = ssi.replace(0E+00, alsi.median())

plt.subplot(1, 2, 2)
plt.imshow(ssi)
plt.title('SSi without gaps', fontweight="bold")

Text(0.5, 1.0, 'SSi without gaps')

```



## ALSI

### Data split

```

x_train = top
y_train = alsi
x_test = bottom
y_test = alsi

```

We are not using XGBRegressor for this problem because size of dataset is large and computation time of XGBRegressor is very long.

```

from numpy import asarray
from sklearn.datasets import make_regression
from keras.models import Sequential
from keras.layers import Dense

model = Sequential()
model.add(Dense(100, input_dim=x_train.shape[1], activation="relu"))
model.add(Dense(32, activation="relu"))
model.add(Dense(y_train.shape[1]))
model.compile(loss="mse", optimizer="adam")
model.summary()

```

Model: "sequential\_12"

Layer (type)	Output Shape	Param #
dense_36 (Dense)	(None, 100)	144100
dense_37 (Dense)	(None, 32)	3232
dense_38 (Dense)	(None, 1440)	47520

```

=====
Total params: 194,852
Trainable params: 194,852
Non-trainable params: 0
=====

```

```

model.fit(x_train, y_train, epochs=1000, batch_size=12, verbose=0)
ypred = model.predict(x_test)

```

```

from sklearn.metrics import mean_squared_error
mean_squared_error(y_test,ypred)

```

0.01585633437039213

For these outputs greyscale images are easy to analyze

```

plt.figure(figsize =(12,12))
plt.subplot(1, 2, 1)
plt.imshow(top)
plt.title('Top half of Mercury Albedo', fontweight="bold")
plt.colorbar(orientation='horizontal', ticks=[])

plt.subplot(1, 2, 2)
plt.imshow(alsi, cmap = 'binary')
plt.title('Mercury Al/Si', fontweight="bold")
plt.colorbar(orientation='horizontal', label= 'Al/Si \n (Low to High)')

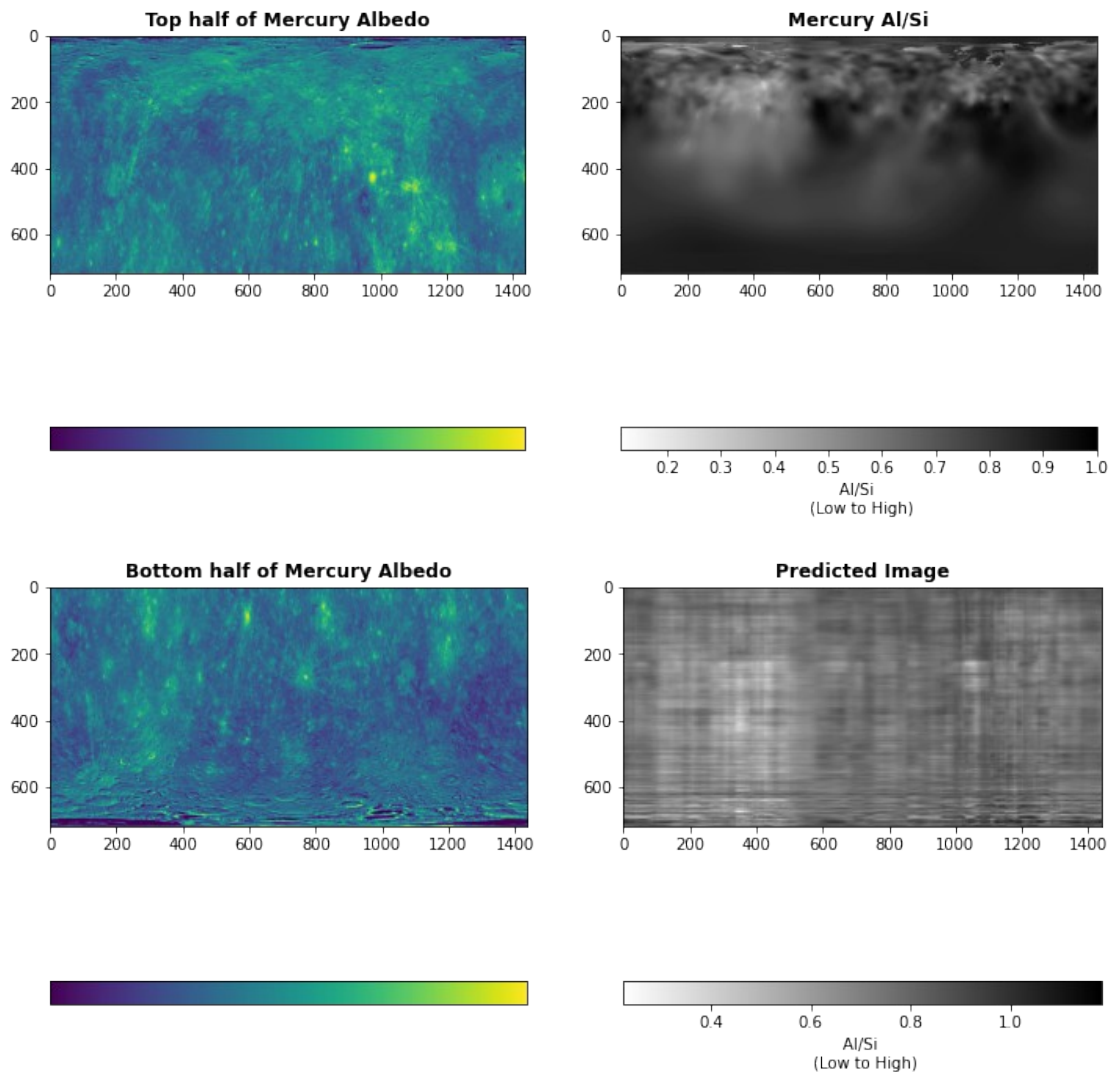
plt.figure(figsize =(12,12))

```

```
plt.subplot(1, 2, 1)
plt.imshow(bottom)
plt.colorbar(orientation='horizontal', ticks=[])
plt.title('Bottom half of Mercury Albedo', fontweight="bold")

plt.subplot(1, 2, 2)
plt.imshow(ypred, cmap = 'binary')
plt.colorbar(orientation='horizontal', label= 'Al/Si \n (Low to High)')
plt.title('Predicted Image', fontweight="bold")

Text(0.5, 1.0, 'Predicted Image')
```



## CASI

```
x_train = top
y_train = casi
x_test = bottom
y_test = casi
```



We are not using XGBRegressor for this problem because size of dataset is large and computation time of XGBRegressor is very long.

```
model = Sequential()
model.add(Dense(100, input_dim=x_train.shape[1], activation="relu"))
model.add(Dense(32, activation="relu"))
model.add(Dense(y_train.shape[1]))
model.compile(loss="mse", optimizer="adam")
model.summary()
```

Model: "sequential\_13"

Layer (type)	Output Shape	Param #
dense_39 (Dense)	(None, 100)	144100
dense_40 (Dense)	(None, 32)	3232
dense_41 (Dense)	(None, 1440)	47520
Total params: 194,852		
Trainable params: 194,852		
Non-trainable params: 0		

```
model.fit(x_train, y_train, epochs=1000, batch_size=12, verbose=0)
ypred = model.predict(x_test)
```

```
mean_squared_error(y_test,ypred)
```

0.02935314307516593

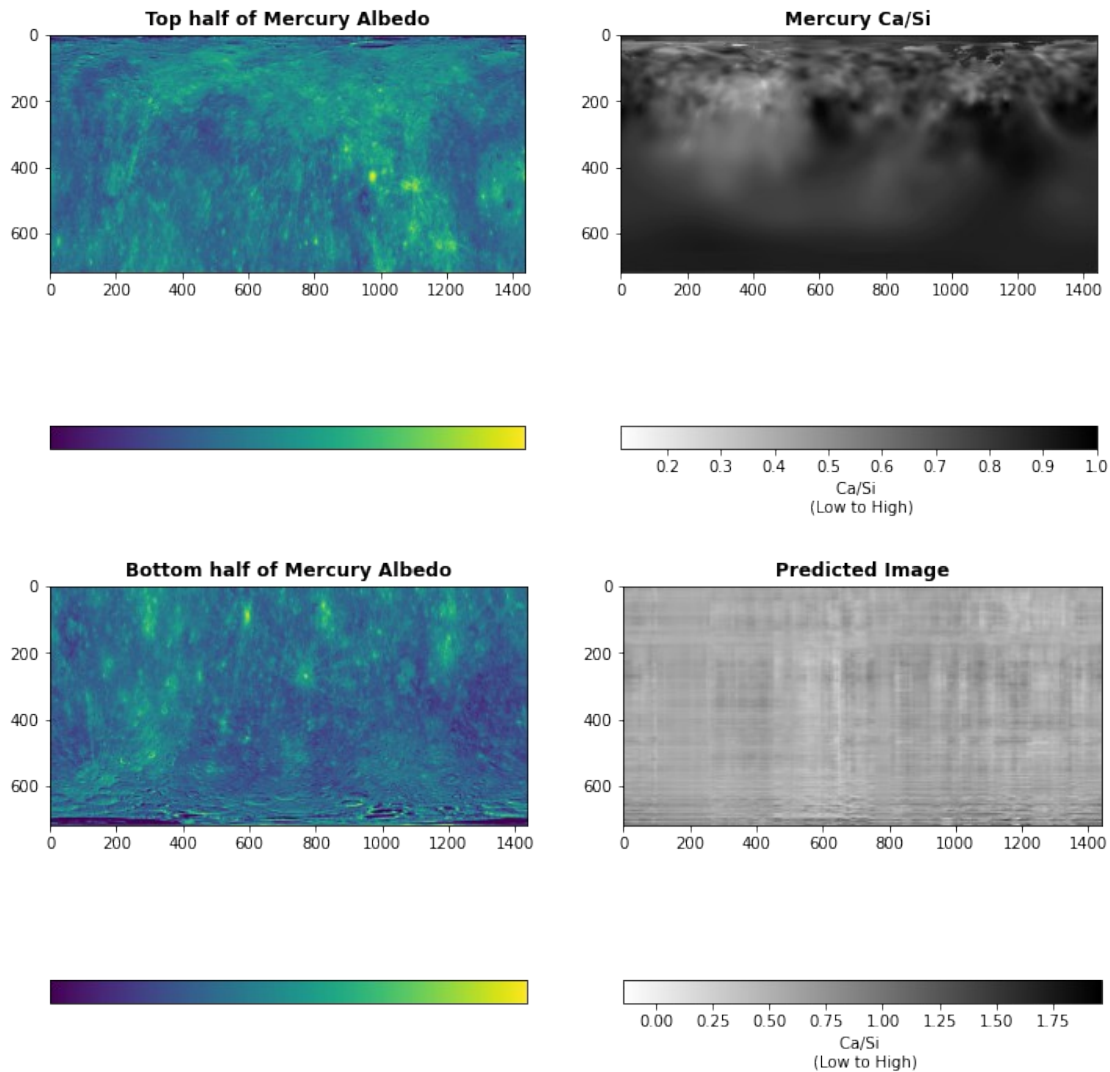
```
plt.figure(figsize =(12,12))
plt.subplot(1, 2, 1)
plt.imshow(top)
plt.title('Top half of Mercury Albedo', fontweight="bold")
plt.colorbar(orientation='horizontal', ticks=[])
```

```
plt.subplot(1, 2, 2)
plt.imshow(alsi, cmap = 'binary')
plt.title('Mercury Ca/Si', fontweight="bold")
plt.colorbar(orientation='horizontal', label= 'Ca/Si \n (Low to High)')
```

```
plt.figure(figsize =(12,12))
plt.subplot(1, 2, 1)
plt.imshow(bottom)
plt.colorbar(orientation='horizontal', ticks=[])
```

```
plt.title('Bottom half of Mercury Albedo', fontweight="bold")
```

```
plt.subplot(1, 2, 2)
plt.imshow(ypred, cmap = 'binary')
plt.colorbar(orientation='horizontal', label= 'Ca/Si \n (Low to High)')
plt.title('Predicted Image', fontweight="bold")
Text(0.5, 1.0, 'Predicted Image')
```



### FESI

```
x_train = top
y_train = fesi
x_test = bottom
y_test = fesi
```

We are not using XGBRegressor for this problem because size of dataset is large and computation time of XGBRegressor is very long.



```

model = Sequential()
model.add(Dense(100, input_dim=x_train.shape[1], activation="relu"))
model.add(Dense(32, activation="relu"))
model.add(Dense(y_train.shape[1]))
model.compile(loss="mse", optimizer="adam")
model.summary()

```

Model: "sequential\_14"

Layer (type)	Output Shape	Param #
dense_42 (Dense)	(None, 100)	144100
dense_43 (Dense)	(None, 32)	3232
dense_44 (Dense)	(None, 1440)	47520

```

=====
Total params: 194,852
Trainable params: 194,852
Non-trainable params: 0
=====

```

```

model.fit(x_train, y_train, epochs=1000, batch_size=12, verbose=0)
ypred = model.predict(x_test)

```

```
mean_squared_error(y_test,ypred)
```

```
0.028232106945608133
```

```

plt.figure(figsize =(12,12))
plt.subplot(1, 2, 1)
plt.imshow(top)
plt.title('Top half of Mercury Albedo', fontweight="bold")
plt.colorbar(orientation='horizontal', ticks=[])

```

```

plt.subplot(1, 2, 2)
plt.imshow(alsi, cmap = 'binary')
plt.title('Mercury Fe/Si', fontweight="bold")
plt.colorbar(orientation='horizontal', label= 'Fe/Si \n (Low to High)')

```

```

plt.figure(figsize =(12,12))
plt.subplot(1, 2, 1)
plt.imshow(bottom)
plt.colorbar(orientation='horizontal', ticks=[])
plt.title('Bottom half of Mercury Albedo', fontweight="bold")

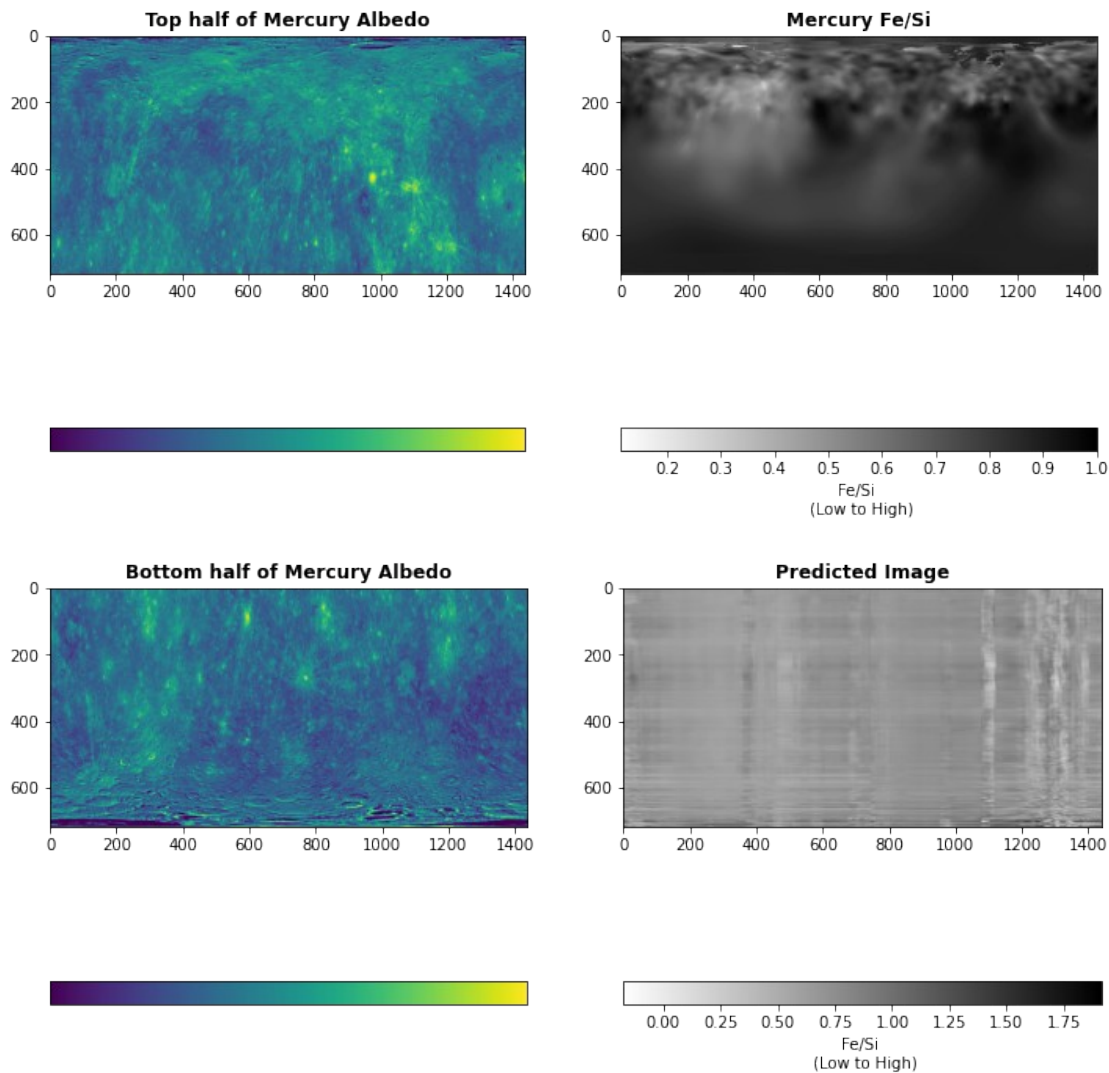
```

```

plt.subplot(1, 2, 2)
plt.imshow(ypred, cmap = 'binary')
plt.colorbar(orientation='horizontal', label= 'Fe/Si \n (Low to

```

```
High)')
plt.title('Predicted Image', fontweight="bold")
Text(0.5, 1.0, 'Predicted Image')
```



## MGSi

```
x_train = top
y_train = mgsi
x_test = bottom
y_test = mgsi
```

We are not using XGBRegressor for this problem because size of dataset is large and computation time of XGBRegressor is very long.

```
model = Sequential()
model.add(Dense(100, input_dim=x_train.shape[1], activation="relu"))
model.add(Dense(32, activation="relu"))
model.add(Dense(y_train.shape[1]))
```

```
model.compile(loss="mse", optimizer="adam")
model.summary()
```

Model: "sequential\_15"

Layer (type)	Output Shape	Param #
dense_45 (Dense)	(None, 100)	144100
dense_46 (Dense)	(None, 32)	3232
dense_47 (Dense)	(None, 1440)	47520

=====  
Total params: 194,852  
Trainable params: 194,852  
Non-trainable params: 0  
=====

```
model.fit(x_train, y_train, epochs=1000, batch_size=12, verbose=0)
ypred = model.predict(x_test)
```

```
mean_squared_error(y_test, ypred)
```

0.012962058376719043

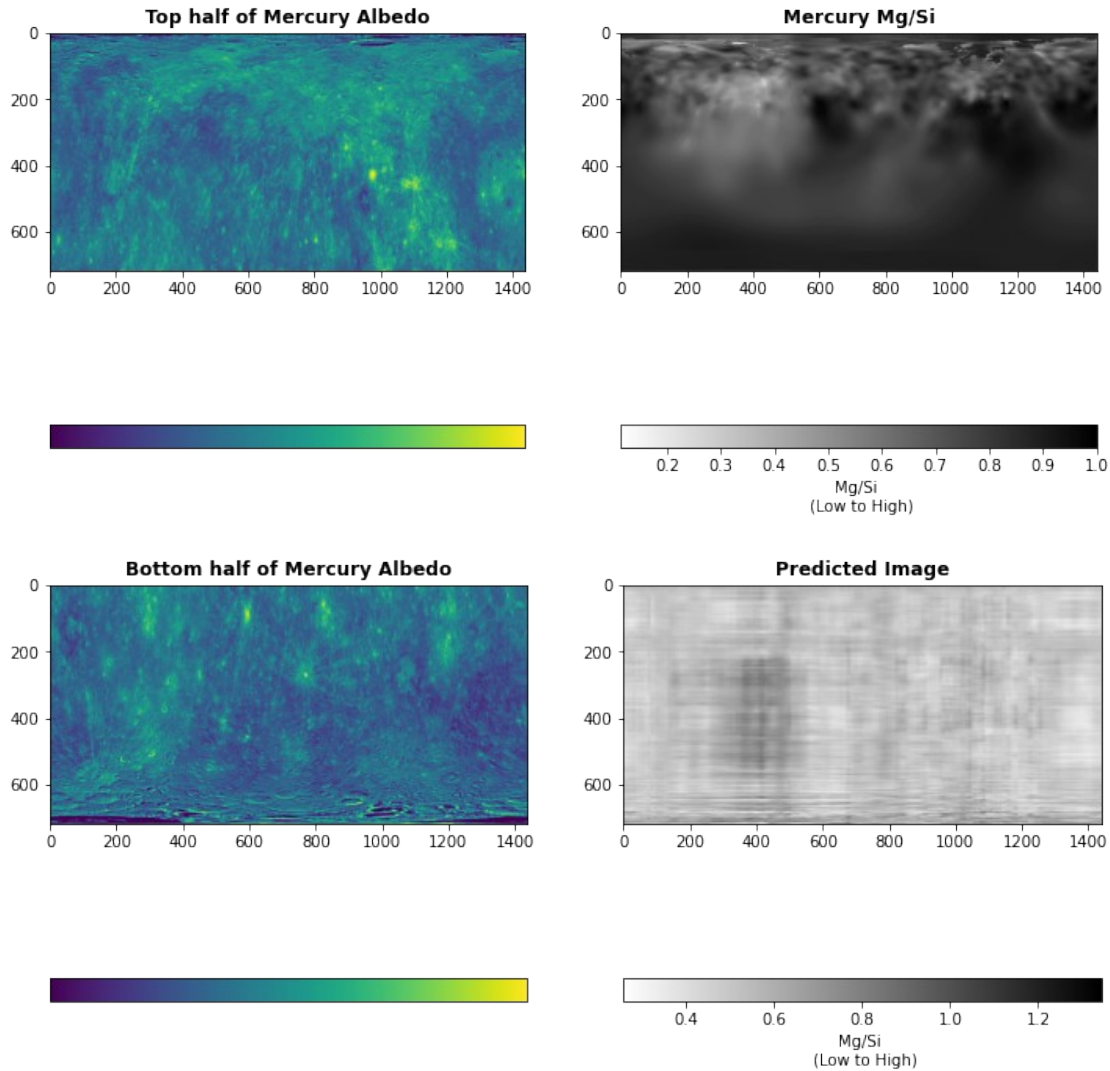
```
plt.figure(figsize=(12,12))
plt.subplot(1, 2, 1)
plt.imshow(top)
plt.title('Top half of Mercury Albedo', fontweight="bold")
plt.colorbar(orientation='horizontal', ticks=[])
```

```
plt.subplot(1, 2, 2)
plt.imshow(alsi, cmap = 'binary')
plt.title('Mercury Mg/Si', fontweight="bold")
plt.colorbar(orientation='horizontal', label= 'Mg/Si \n (Low to High)')
```

```
plt.figure(figsize=(12,12))
plt.subplot(1, 2, 1)
plt.imshow(bottom)
plt.colorbar(orientation='horizontal', ticks=[])
plt.title('Bottom half of Mercury Albedo', fontweight="bold")
```

```
plt.subplot(1, 2, 2)
plt.imshow(ypred, cmap = 'binary')
plt.colorbar(orientation='horizontal', label= 'Mg/Si \n (Low to High)')
plt.title('Predicted Image', fontweight="bold")
```

```
Text(0.5, 1.0, 'Predicted Image')
```



## SSI

```
x_train = top
y_train = ssi
x_test = bottom
y_test = ssi
```

We are not using XGBRegressor for this problem because size of dataset is large and computation time of XGBRegressor is very long.

```
model = Sequential()
model.add(Dense(100, input_dim=x_train.shape[1], activation="relu"))
model.add(Dense(32, activation="relu"))
model.add(Dense(y_train.shape[1]))
model.compile(loss="mse", optimizer="adam")
model.summary()
```

Model: "sequential\_16"

---

Layer (type)	Output Shape	Param #
dense_48 (Dense)	(None, 100)	144100
dense_49 (Dense)	(None, 32)	3232
dense_50 (Dense)	(None, 1440)	47520
=====		
Total params: 194,852		
Trainable params: 194,852		
Non-trainable params: 0		

```
model.fit(x_train, y_train, epochs=1000, batch_size=12, verbose=0)
ypred = model.predict(x_test)
```

```
mean_squared_error(y_test, ypred)
```

```
0.040603376142447495
```

```
plt.figure(figsize=(12,12))
plt.subplot(1, 2, 1)
plt.imshow(top)
plt.title('Top half of Mercury Albedo', fontweight="bold")
plt.colorbar(orientation='horizontal', ticks=[])

plt.subplot(1, 2, 2)
plt.imshow(alsi, cmap = 'binary')
plt.title('Mercury S/Si', fontweight="bold")
plt.colorbar(orientation='horizontal', label= 'S/Si \n (Low to High)')

plt.figure(figsize=(12,12))
plt.subplot(1, 2, 1)
plt.imshow(bottom)
plt.colorbar(orientation='horizontal', ticks=[])
plt.title('Bottom half of Mercury Albedo', fontweight="bold")

plt.subplot(1, 2, 2)
plt.imshow(ypred, cmap = 'binary')
plt.colorbar(orientation='horizontal', label= 'S/Si \n (Low to High)')
plt.title('Predicted Image', fontweight="bold")

Text(0.5, 1.0, 'Predicted Image')
```

