

# final

November 18, 2018

```
In [2]: # dataset - Sloan Digital Sky Survey RD14
# Number of rows - 600
#Number of columns - 18
# number of categorical features - 7
#number of numerical features - 9
```

```
%matplotlib inline
import numpy as np
import pandas as pd
import pylab as pl
import matplotlib.pyplot as plt
from sklearn import preprocessing
import scipy.stats
```

```
In [3]: df = pd.read_csv('sky1.csv')
```

```
In [4]: df
```

```
Out[4]:
```

	objid	ra	dec	u	g	r	\
0	1.240000e+18	183.531326	0.089693	19.47406	17.04240	15.94699	
1	1.240000e+18	183.598371	0.135285	18.66280	17.21449	16.67637	
2	1.240000e+18	183.680207	0.126185	19.38298	18.19169	17.47428	
3	1.240000e+18	183.870529	0.049911	17.76536	16.60272	16.16116	
4	1.240000e+18	183.883288	0.102557	17.55025	16.26342	16.43869	
5	1.240000e+18	183.847174	0.173694	19.43133	18.46779	18.16451	
6	1.240000e+18	183.864379	0.019201	19.38322	17.88995	17.10537	
7	1.240000e+18	183.900081	0.187473	18.97993	17.84496	17.38022	
8	1.240000e+18	183.924588	0.097246	17.90616	16.97172	16.67541	
9	1.240000e+18	183.973498	0.081626	18.67249	17.71375	17.49362	
10	1.240000e+18	183.979195	0.135998	19.29772	17.80227	17.18266	
11	1.240000e+18	184.085331	0.112110	18.83307	17.51785	16.94273	
12	1.240000e+18	184.102098	0.191511	19.56250	18.19113	17.65759	
13	1.240000e+18	184.160510	0.075645	19.57990	17.72815	16.98740	
14	1.240000e+18	184.189574	0.099482	19.25667	17.54869	16.63578	
15	1.240000e+18	184.350647	0.207230	18.73832	18.60962	18.39696	
16	1.240000e+18	184.221797	0.046955	19.05958	18.09512	17.92766	
17	1.240000e+18	184.245664	0.198257	19.22143	19.30248	19.13823	
18	1.240000e+18	184.388870	0.068287	19.04397	17.51106	16.87335	

19	1.240000e+18	184.380919	0.174323	17.81661	16.86976	16.53884
20	1.240000e+18	184.466853	0.111965	19.39320	18.48274	18.16551
21	1.240000e+18	184.569411	0.137091	17.51339	16.41793	16.06695
22	1.240000e+18	184.654170	0.122673	19.07731	18.64518	18.49678
23	1.240000e+18	184.658795	0.159936	18.63032	17.23437	16.72749
24	1.240000e+18	184.516834	0.145550	19.21684	18.29956	18.19589
25	1.240000e+18	184.618563	0.080576	19.41270	17.47109	16.49870
26	1.240000e+18	184.627408	0.013234	18.48543	17.33900	16.90179
27	1.240000e+18	184.676762	0.059442	17.72835	16.19783	15.66073
28	1.240000e+18	184.711970	0.066572	17.70824	16.61069	16.19772
29	1.240000e+18	184.770204	0.198010	18.18974	17.23738	16.89113
..	...	...	...	...	...	...
569	1.240000e+18	244.284910	-0.455477	18.63084	16.69326	15.67451
570	1.240000e+18	244.491831	-0.470566	18.43597	17.47617	17.10257
571	1.240000e+18	244.496791	-0.498398	17.98780	16.51811	15.74451
572	1.240000e+18	244.916397	-0.523076	18.45084	17.39167	16.97855
573	1.240000e+18	244.925582	-0.455836	19.34707	17.76923	16.92714
574	1.240000e+18	244.935545	-0.479247	18.53797	17.16947	16.50210
575	1.240000e+18	245.128493	-0.499415	19.18215	18.79711	18.81150
576	1.240000e+18	245.367353	-0.457074	18.12316	16.31978	15.37991
577	1.240000e+18	245.376694	-0.457008	19.46922	17.96987	17.31625
578	1.240000e+18	245.978479	-0.483455	19.38072	17.38953	16.26663
579	1.240000e+18	246.034378	-0.473284	19.51987	17.56103	16.44300
580	1.240000e+18	246.268157	-0.426990	19.46025	18.28482	17.78494
581	1.240000e+18	246.695626	-0.510111	18.01060	16.81185	16.32209
582	1.240000e+18	246.782081	-0.492432	17.60994	15.90911	15.02090
583	1.240000e+18	247.021139	-0.448677	19.23512	19.08207	18.78046
584	1.240000e+18	248.269016	-0.488920	19.27635	17.99304	17.47953
585	1.240000e+18	248.894520	-0.452329	18.73576	16.83826	15.94847
586	1.240000e+18	249.496551	-0.498468	18.99369	17.31122	16.48666
587	1.240000e+18	189.429821	-0.131042	17.83883	16.62422	16.84056
588	1.240000e+18	189.448544	-0.052745	18.97562	17.84829	17.49412
589	1.240000e+18	189.453801	-0.097313	17.66081	16.62739	16.28842
590	1.240000e+18	189.468747	-0.036000	17.28231	16.20900	15.76480
591	1.240000e+18	189.522249	-0.027031	16.32175	14.72385	13.88480
592	1.240000e+18	189.497150	-0.110585	19.25978	18.10524	17.42960
593	1.240000e+18	225.529253	-0.029788	19.21136	18.73800	18.76889
594	1.240000e+18	225.494356	-0.114317	19.43974	18.09289	17.40579
595	1.240000e+18	145.288576	0.346512	19.22352	18.00232	17.37943
596	1.240000e+18	145.292821	0.376326	19.31302	18.13527	17.67289
597	1.240000e+18	145.593807	0.278308	19.10790	18.92201	18.59314
598	1.240000e+18	145.637275	0.373311	19.20753	17.87736	17.30583

	i	z	run	rerun	camcol	field	specobjid	class \
0	15.50342	15.22531	752.0	301.0	4	267	3.720000e+18	STAR
1	NaN	16.39150	752.0	301.0	4	267	3.640000e+17	STAR
2	17.08732	16.80125	752.0	301.0	4	268	3.230000e+17	GALAXY
3	15.98233	15.90438	752.0	301.0	4	269	3.720000e+18	STAR

4	NaN	16.61326	752.0	NaN	4	269	3.720000e+18	NaN
5	18.01475	18.04155	752.0	301.0	4	269	3.650000e+17	STAR
6	16.66393	16.36955	752.0	301.0	4	269	3.230000e+17	GALAXY
7	17.20673	17.07071	NaN	301.0	4	269	3.720000e+18	STAR
8	NaN	16.47596	752.0	301.0	4	270	3.640000e+17	STAR
9	17.28284	17.22644	752.0	301.0	4	270	3.240000e+17	GALAXY
10	16.92335	16.79928	752.0	301.0	4	270	3.720000e+18	STAR
11	16.71418	16.60521	752.0	301.0	4	271	3.720000e+18	STAR
12	17.47573	17.39203	752.0	301.0	4	271	3.720000e+18	STAR
13	16.68076	16.50426	752.0	301.0	4	271	3.720000e+18	STAR
14	16.14922	15.76639	752.0	301.0	4	271	3.240000e+17	GALAXY
15	18.31174	17.97663	752.0	301.0	4	272	3.230000e+17	QSO
16	NaN	17.90772	752.0	301.0	4	272	3.650000e+17	STAR
17	19.11351	19.23454	752.0	301.0	4	272	3.230000e+17	QSO
18	16.61114	16.48303	752.0	301.0	4	273	3.640000e+17	STAR
19	16.19576	16.08668	752.0	301.0	4	273	3.230000e+17	GALAXY
20	18.05122	18.04328	752.0	301.0	4	273	3.650000e+17	STAR
21	15.93751	15.89478	752.0	301.0	4	274	3.240000e+17	STAR
22	18.52677	18.45765	752.0	301.0	4	274	3.240000e+17	QSO
23	16.51772	16.44039	752.0	301.0	4	274	2.880000e+18	STAR
24	18.14758	18.26658	752.0	301.0	4	274	3.720000e+18	STAR
25	16.06502	15.66848	752.0	301.0	4	274	3.240000e+17	GALAXY
26	16.66712	16.48684	752.0	301.0	4	274	3.230000e+17	NaN
27	15.51586	15.43615	752.0	301.0	4	275	2.880000e+18	STAR
28	NaN	16.00229	752.0	301.0	4	275	3.720000e+18	STAR
29	16.75727	16.71052	752.0	301.0	4	275	2.880000e+18	STAR
..	...	...	...	...	...	...	...	...
569	15.20557	14.82077	745.0	301.0	2	571	3.900000e+17	GALAXY
570	16.90971	16.82976	745.0	301.0	2	573	3.900000e+17	GALAXY
571	15.30433	14.98556	745.0	301.0	2	573	3.900000e+17	GALAXY
572	16.82599	16.76456	745.0	301.0	2	576	3.900000e+17	STAR
573	16.48486	16.14608	745.0	301.0	2	576	3.900000e+17	GALAXY
574	16.18933	15.95892	745.0	301.0	2	576	3.900000e+17	GALAXY
575	18.61492	18.51464	745.0	301.0	2	577	4.100000e+17	QSO
576	14.91012	14.53670	745.0	301.0	2	579	4.100000e+17	GALAXY
577	16.92729	16.71774	745.0	301.0	2	579	3.900000e+17	GALAXY
578	15.75659	15.36963	745.0	301.0	2	583	4.100000e+17	NaN
579	15.95302	15.57304	745.0	301.0	2	583	4.100000e+17	GALAXY
580	17.50377	17.31497	745.0	301.0	2	585	4.100000e+17	GALAXY
581	16.03653	15.92627	745.0	301.0	2	588	4.100000e+17	GALAXY
582	14.54955	14.19971	745.0	301.0	2	588	4.100000e+17	GALAXY
583	18.81124	18.99055	745.0	301.0	2	590	4.100000e+17	QSO
584	17.29088	18.28560	745.0	301.0	2	598	3.920000e+17	STAR
585	15.53803	15.15638	745.0	301.0	2	602	3.920000e+17	GALAXY
586	16.05332	15.74858	745.0	301.0	2	606	3.920000e+17	GALAXY
587	17.03584	17.15817	745.0	301.0	3	205	3.260000e+18	STAR
588	17.37400	17.35097	745.0	301.0	3	205	3.280000e+17	STAR
589	16.17215	16.14751	745.0	301.0	3	205	3.260000e+18	STAR

590	15.51375	15.47968	745.0	301.0	3	205	3.260000e+18	STAR
591	13.46923	13.15061	745.0	301.0	3	205	3.270000e+17	GALAXY
592	17.08964	16.85153	745.0	301.0	3	205	3.280000e+17	GALAXY
593	18.96195	19.18182	745.0	301.0	3	446	4.520000e+18	STAR
594	17.00810	16.82082	745.0	301.0	3	446	3.490000e+17	GALAXY
595	17.01110	16.76524	756.0	301.0	4	197	3.000000e+17	GALAXY
596	17.49996	17.44187	756.0	301.0	4	197	3.000000e+17	STAR
597	18.47573	18.40461	756.0	301.0	4	199	3.000000e+17	QSO
598	17.08623	16.96132	756.0	301.0	4	199	3.000000e+17	STAR

	redshift	plate	mjd	fiberid
0	-8.960000e-06	3306	54922	491
1	-5.490000e-05	323	51615	541
2	1.231112e-01	287	52023	513
3	-1.106160e-04	3306	54922	510
4	5.903570e-04	3306	54922	512
5	3.146030e-04	324	51666	594
6	1.002423e-01	287	52023	559
7	3.148480e-04	3306	54922	515
8	8.910000e-05	323	51615	595
9	4.050813e-02	288	52000	400
10	-3.460000e-05	3306	54922	506
11	6.227740e-04	3306	54922	547
12	5.470000e-05	3306	54922	544
13	8.300000e-06	3306	54922	546
14	7.208736e-02	288	52000	389
15	2.719369e-01	287	52023	587
16	2.099740e-04	324	51666	35
17	1.178098e+00	287	52023	583
18	2.400980e-04	323	51615	628
19	7.277206e-02	287	52023	632
20	-2.702180e-04	324	51666	623
21	-3.550000e-05	288	52000	430
22	9.251733e-01	288	52000	421
23	4.310000e-06	2558	54140	356
24	8.800000e-05	3306	54922	595
25	1.167829e-01	288	52000	437
26	2.129653e-02	287	52023	637
27	1.753650e-04	2558	54140	358
28	1.723800e-04	3306	54922	632
29	5.198080e-04	2558	54140	349
..	...	...	...	...
569	9.280799e-02	346	51693	162
570	1.660742e-02	346	51693	148
571	5.047690e-02	346	51693	146
572	-3.407500e-04	346	51693	81
573	7.900908e-02	346	51693	106
574	5.936759e-02	346	51693	108

575	5.382954e-01	364	52000	307
576	5.909181e-02	364	52000	302
577	6.916405e-02	346	51693	67
578	9.362785e-02	364	52000	221
579	1.424379e-01	364	52000	190
580	4.653732e-02	364	52000	161
581	4.449500e-02	364	52000	146
582	4.619161e-02	364	52000	107
583	1.149604e+00	364	52000	74
584	-7.320000e-05	348	51671	236
585	7.018682e-02	348	51671	163
586	5.894065e-02	348	51671	106
587	6.484750e-04	2895	54567	116
588	2.573870e-04	291	51928	310
589	5.910000e-07	2895	54567	117
590	5.791130e-04	2895	54567	119
591	1.251297e-02	290	51941	516
592	1.383053e-01	291	51928	318
593	-1.976670e-04	4016	55632	396
594	4.019753e-02	310	51990	224
595	6.460119e-02	266	51630	359
596	2.133720e-04	266	51630	346
597	1.404700e+00	266	51630	396
598	-3.480000e-05	266	51630	390

[599 rows x 18 columns]

```
In [5]: #To count the number of NAN values in each column
        #NAN value in class column(categorical feature)
        len(df) - df['class'].count()
```

Out[5]: 20

```
In [6]: #NAN values in i column or feature (numerical feature)
        len(df) - df['i'].count()
```

Out[6]: 11

```
In [7]: # to find the number of NAN values of every feature
        len(df) - df.count()
```

```
Out[7]: objid      0
        ra         0
        dec        0
        u          0
        g          0
        r          0
        i         11
        z          0
```

```

run          9
rerun        9
camcol       0
field        0
specobjid    0
class        20
redshift     0
plate        0
mjd          0
fiberid      0
dtype: int64

```

```
In [8]: # 1) Data Cleaning
```

```
In [9]: # All the NAN's for categorical columns replaced with its previous row value
```

```
In [10]: df['class'].ffill(inplace=True)
```

```
In [11]: df['run'].ffill(inplace=True)
```

```
In [12]: df['rerun'].ffill(inplace=True)
```

```
In [13]: # All the NAN's for numeric columns replaced with average of the column
```

```
In [14]: df['i'].fillna(df['i'].mean(),inplace=True)
```

```
In [15]: # Number of NAN values after filling with respective values
```

```
In [16]: len(df) - df.count()
```

```

Out[16]: objid      0
         ra         0
         dec        0
         u          0
         g          0
         r          0
         i          0
         z          0
         run        0
         rerun      0
         camcol     0
         field      0
         specobjid  0
         class      0
         redshift   0
         plate      0
         mjd        0
         fiberid    0
         dtype: int64

```

```

In [17]: # now our data is complete with no NAN values

In [18]: # By studying the data we observe that objid and rerun values are the same for every row

In [19]: del df['objid']
          del df['rerun']

In [20]: # our dataset after Data Cleaning process

In [21]: df.to_csv('cleanedsky.csv', encoding='utf-8', index=False)
          df

```

```

Out[21]:

```

	ra	dec	u	g	r	i	z	\
0	183.531326	0.089693	19.47406	17.04240	15.94699	15.503420	15.22531	
1	183.598371	0.135285	18.66280	17.21449	16.67637	16.544191	16.39150	
2	183.680207	0.126185	19.38298	18.19169	17.47428	17.087320	16.80125	
3	183.870529	0.049911	17.76536	16.60272	16.16116	15.982330	15.90438	
4	183.883288	0.102557	17.55025	16.26342	16.43869	16.544191	16.61326	
5	183.847174	0.173694	19.43133	18.46779	18.16451	18.014750	18.04155	
6	183.864379	0.019201	19.38322	17.88995	17.10537	16.663930	16.36955	
7	183.900081	0.187473	18.97993	17.84496	17.38022	17.206730	17.07071	
8	183.924588	0.097246	17.90616	16.97172	16.67541	16.544191	16.47596	
9	183.973498	0.081626	18.67249	17.71375	17.49362	17.282840	17.22644	
10	183.979195	0.135998	19.29772	17.80227	17.18266	16.923350	16.79928	
11	184.085331	0.112110	18.83307	17.51785	16.94273	16.714180	16.60521	
12	184.102098	0.191511	19.56250	18.19113	17.65759	17.475730	17.39203	
13	184.160510	0.075645	19.57990	17.72815	16.98740	16.680760	16.50426	
14	184.189574	0.099482	19.25667	17.54869	16.63578	16.149220	15.76639	
15	184.350647	0.207230	18.73832	18.60962	18.39696	18.311740	17.97663	
16	184.221797	0.046955	19.05958	18.09512	17.92766	16.544191	17.90772	
17	184.245664	0.198257	19.22143	19.30248	19.13823	19.113510	19.23454	
18	184.388870	0.068287	19.04397	17.51106	16.87335	16.611140	16.48303	
19	184.380919	0.174323	17.81661	16.86976	16.53884	16.195760	16.08668	
20	184.466853	0.111965	19.39320	18.48274	18.16551	18.051220	18.04328	
21	184.569411	0.137091	17.51339	16.41793	16.06695	15.937510	15.89478	
22	184.654170	0.122673	19.07731	18.64518	18.49678	18.526770	18.45765	
23	184.658795	0.159936	18.63032	17.23437	16.72749	16.517720	16.44039	
24	184.516834	0.145550	19.21684	18.29956	18.19589	18.147580	18.26658	
25	184.618563	0.080576	19.41270	17.47109	16.49870	16.065020	15.66848	
26	184.627408	0.013234	18.48543	17.33900	16.90179	16.667120	16.48684	
27	184.676762	0.059442	17.72835	16.19783	15.66073	15.515860	15.43615	
28	184.711970	0.066572	17.70824	16.61069	16.19772	16.544191	16.00229	
29	184.770204	0.198010	18.18974	17.23738	16.89113	16.757270	16.71052	
...	...	...	...	...	...	...	...	
569	244.284910	-0.455477	18.63084	16.69326	15.67451	15.205570	14.82077	
570	244.491831	-0.470566	18.43597	17.47617	17.10257	16.909710	16.82976	
571	244.496791	-0.498398	17.98780	16.51811	15.74451	15.304330	14.98556	
572	244.916397	-0.523076	18.45084	17.39167	16.97855	16.825990	16.76456	
573	244.925582	-0.455836	19.34707	17.76923	16.92714	16.484860	16.14608	

574	244.935545	-0.479247	18.53797	17.16947	16.50210	16.189330	15.95892
575	245.128493	-0.499415	19.18215	18.79711	18.81150	18.614920	18.51464
576	245.367353	-0.457074	18.12316	16.31978	15.37991	14.910120	14.53670
577	245.376694	-0.457008	19.46922	17.96987	17.31625	16.927290	16.71774
578	245.978479	-0.483455	19.38072	17.38953	16.26663	15.756590	15.36963
579	246.034378	-0.473284	19.51987	17.56103	16.44300	15.953020	15.57304
580	246.268157	-0.426990	19.46025	18.28482	17.78494	17.503770	17.31497
581	246.695626	-0.510111	18.01060	16.81185	16.32209	16.036530	15.92627
582	246.782081	-0.492432	17.60994	15.90911	15.02090	14.549550	14.19971
583	247.021139	-0.448677	19.23512	19.08207	18.78046	18.811240	18.99055
584	248.269016	-0.488920	19.27635	17.99304	17.47953	17.290880	18.28560
585	248.894520	-0.452329	18.73576	16.83826	15.94847	15.538030	15.15638
586	249.496551	-0.498468	18.99369	17.31122	16.48666	16.053320	15.74858
587	189.429821	-0.131042	17.83883	16.62422	16.84056	17.035840	17.15817
588	189.448544	-0.052745	18.97562	17.84829	17.49412	17.374000	17.35097
589	189.453801	-0.097313	17.66081	16.62739	16.28842	16.172150	16.14751
590	189.468747	-0.036000	17.28231	16.20900	15.76480	15.513750	15.47968
591	189.522249	-0.027031	16.32175	14.72385	13.88480	13.469230	13.15061
592	189.497150	-0.110585	19.25978	18.10524	17.42960	17.089640	16.85153
593	225.529253	-0.029788	19.21136	18.73800	18.76889	18.961950	19.18182
594	225.494356	-0.114317	19.43974	18.09289	17.40579	17.008100	16.82082
595	145.288576	0.346512	19.22352	18.00232	17.37943	17.011100	16.76524
596	145.292821	0.376326	19.31302	18.13527	17.67289	17.499960	17.44187
597	145.593807	0.278308	19.10790	18.92201	18.59314	18.475730	18.40461
598	145.637275	0.373311	19.20753	17.87736	17.30583	17.086230	16.96132

	run	camcol	field	specobjid	class	redshift	plate	mjd \
0	752.0	4	267	3.720000e+18	STAR	-8.960000e-06	3306	54922
1	752.0	4	267	3.640000e+17	STAR	-5.490000e-05	323	51615
2	752.0	4	268	3.230000e+17	GALAXY	1.231112e-01	287	52023
3	752.0	4	269	3.720000e+18	STAR	-1.106160e-04	3306	54922
4	752.0	4	269	3.720000e+18	STAR	5.903570e-04	3306	54922
5	752.0	4	269	3.650000e+17	STAR	3.146030e-04	324	51666
6	752.0	4	269	3.230000e+17	GALAXY	1.002423e-01	287	52023
7	752.0	4	269	3.720000e+18	STAR	3.148480e-04	3306	54922
8	752.0	4	270	3.640000e+17	STAR	8.910000e-05	323	51615
9	752.0	4	270	3.240000e+17	GALAXY	4.050813e-02	288	52000
10	752.0	4	270	3.720000e+18	STAR	-3.460000e-05	3306	54922
11	752.0	4	271	3.720000e+18	STAR	6.227740e-04	3306	54922
12	752.0	4	271	3.720000e+18	STAR	5.470000e-05	3306	54922
13	752.0	4	271	3.720000e+18	STAR	8.300000e-06	3306	54922
14	752.0	4	271	3.240000e+17	GALAXY	7.208736e-02	288	52000
15	752.0	4	272	3.230000e+17	QSO	2.719369e-01	287	52023
16	752.0	4	272	3.650000e+17	STAR	2.099740e-04	324	51666
17	752.0	4	272	3.230000e+17	QSO	1.178098e+00	287	52023
18	752.0	4	273	3.640000e+17	STAR	2.400980e-04	323	51615
19	752.0	4	273	3.230000e+17	GALAXY	7.277206e-02	287	52023
20	752.0	4	273	3.650000e+17	STAR	-2.702180e-04	324	51666



21	752.0	4	274	3.240000e+17	STAR	-3.550000e-05	288	52000
22	752.0	4	274	3.240000e+17	QSO	9.251733e-01	288	52000
23	752.0	4	274	2.880000e+18	STAR	4.310000e-06	2558	54140
24	752.0	4	274	3.720000e+18	STAR	8.800000e-05	3306	54922
25	752.0	4	274	3.240000e+17	GALAXY	1.167829e-01	288	52000
26	752.0	4	274	3.230000e+17	GALAXY	2.129653e-02	287	52023
27	752.0	4	275	2.880000e+18	STAR	1.753650e-04	2558	54140
28	752.0	4	275	3.720000e+18	STAR	1.723800e-04	3306	54922
29	752.0	4	275	2.880000e+18	STAR	5.198080e-04	2558	54140
..	...	...	...	...	...	...	...	...
569	745.0	2	571	3.900000e+17	GALAXY	9.280799e-02	346	51693
570	745.0	2	573	3.900000e+17	GALAXY	1.660742e-02	346	51693
571	745.0	2	573	3.900000e+17	GALAXY	5.047690e-02	346	51693
572	745.0	2	576	3.900000e+17	STAR	-3.407500e-04	346	51693
573	745.0	2	576	3.900000e+17	GALAXY	7.900908e-02	346	51693
574	745.0	2	576	3.900000e+17	GALAXY	5.936759e-02	346	51693
575	745.0	2	577	4.100000e+17	QSO	5.382954e-01	364	52000
576	745.0	2	579	4.100000e+17	GALAXY	5.909181e-02	364	52000
577	745.0	2	579	3.900000e+17	GALAXY	6.916405e-02	346	51693
578	745.0	2	583	4.100000e+17	GALAXY	9.362785e-02	364	52000
579	745.0	2	583	4.100000e+17	GALAXY	1.424379e-01	364	52000
580	745.0	2	585	4.100000e+17	GALAXY	4.653732e-02	364	52000
581	745.0	2	588	4.100000e+17	GALAXY	4.449500e-02	364	52000
582	745.0	2	588	4.100000e+17	GALAXY	4.619161e-02	364	52000
583	745.0	2	590	4.100000e+17	QSO	1.149604e+00	364	52000
584	745.0	2	598	3.920000e+17	STAR	-7.320000e-05	348	51671
585	745.0	2	602	3.920000e+17	GALAXY	7.018682e-02	348	51671
586	745.0	2	606	3.920000e+17	GALAXY	5.894065e-02	348	51671
587	745.0	3	205	3.260000e+18	STAR	6.484750e-04	2895	54567
588	745.0	3	205	3.280000e+17	STAR	2.573870e-04	291	51928
589	745.0	3	205	3.260000e+18	STAR	5.910000e-07	2895	54567
590	745.0	3	205	3.260000e+18	STAR	5.791130e-04	2895	54567
591	745.0	3	205	3.270000e+17	GALAXY	1.251297e-02	290	51941
592	745.0	3	205	3.280000e+17	GALAXY	1.383053e-01	291	51928
593	745.0	3	446	4.520000e+18	STAR	-1.976670e-04	4016	55632
594	745.0	3	446	3.490000e+17	GALAXY	4.019753e-02	310	51990
595	756.0	4	197	3.000000e+17	GALAXY	6.460119e-02	266	51630
596	756.0	4	197	3.000000e+17	STAR	2.133720e-04	266	51630
597	756.0	4	199	3.000000e+17	QSO	1.404700e+00	266	51630
598	756.0	4	199	3.000000e+17	STAR	-3.480000e-05	266	51630

	fiberid
0	491
1	541
2	513
3	510
4	512
5	594

6	559
7	515
8	595
9	400
10	506
11	547
12	544
13	546
14	389
15	587
16	35
17	583
18	628
19	632
20	623
21	430
22	421
23	356
24	595
25	437
26	637
27	358
28	632
29	349
..	...
569	162
570	148
571	146
572	81
573	106
574	108
575	307
576	302
577	67
578	221
579	190
580	161
581	146
582	107
583	74
584	236
585	163
586	106
587	116
588	310
589	117
590	119
591	516

```

592      318
593      396
594      224
595      359
596      346
597      396
598      390

```

```
[599 rows x 16 columns]
```

```
In [22]: # 2) Normalization and Standardization
```

```
In [23]: # Why is normalization important?
```

```

''' The normal distribution is important because of the Central limit theorem. In simple
words, the normal distribution is the most common distribution of data.

Another corollary is that the normal distribution makes math easy - things like calcula

```

```
Out[23]: " The normal distribution is important because of the Central limit theorem. In simple
```

```
In [24]: # How does it affect dataset?
```

```

'''Normalization is performed on data to remove amplitude variation and only focus on t
he shape of the distribution.

''' normalization brings any dataset to a comparable range. It could be to squash down th

```

```
Out[24]: ' normalization brings any dataset to a comparable range. It could be to squash down th
```

```
In [25]: # Different graphs used to check whether the data is normal
```

```

# Histograms
# boxplots
# normal probability plot or qq plot or quantile-quantile plot

```

```
In [26]: # Normalizing all the numeric columns, to make mean 0 and variance 1
```

```
In [27]: # Normalizing only numeric features
```

```

cols_to_norm = ['ra', 'dec', 'u', 'g', 'r', 'i', 'z', 'redshift', 'mjd']
df[cols_to_norm] = df[cols_to_norm].apply(lambda x: (x - x.min()) / (x.max() - x.min()))

```

```
In [28]: df
```

```

Out[28]:
      ra      dec      u      g      r      i      z  \
0  0.526143  0.501836  0.973617  0.548873  0.340158  0.186836  0.290166
1  0.526624  0.518848  0.791571  0.579327  0.460469  0.282428  0.453268
2  0.527210  0.515452  0.953179  0.752260  0.592084  0.332313  0.510575
3  0.528573  0.486993  0.590185  0.471064  0.375485  0.230822  0.385140
4  0.528664  0.506636  0.541915  0.411019  0.421264  0.282428  0.484283
5  0.528406  0.533179  0.964029  0.801121  0.705937  0.417496  0.684043
6  0.528529  0.475534  0.953233  0.698862  0.531232  0.293426  0.450198
7  0.528785  0.538321  0.862735  0.690900  0.576569  0.343281  0.548262
8  0.528960  0.504654  0.621781  0.536365  0.460311  0.282428  0.465081

```

9	0.529310	0.498826	0.793745	0.667680	0.595274	0.350271	0.570042
10	0.529351	0.519114	0.934047	0.683345	0.543981	0.317253	0.510300
11	0.530111	0.510201	0.829779	0.633012	0.504405	0.298041	0.483157
12	0.530231	0.539827	0.993463	0.752161	0.622321	0.367988	0.593201
13	0.530650	0.496595	0.997368	0.670228	0.511773	0.294972	0.469039
14	0.530858	0.505489	0.924835	0.638470	0.453774	0.246151	0.365841
15	0.532012	0.545692	0.808517	0.826220	0.744280	0.444773	0.674963
16	0.531089	0.485890	0.880608	0.735170	0.666869	0.282428	0.665325
17	0.531260	0.542344	0.916927	0.948833	0.866552	0.518414	0.850893
18	0.532285	0.493849	0.877105	0.631811	0.492961	0.288577	0.466069
19	0.532228	0.533414	0.601686	0.518321	0.437783	0.250425	0.410636
20	0.532844	0.510147	0.955472	0.803766	0.706102	0.420845	0.684285
21	0.533578	0.519522	0.533643	0.438362	0.359945	0.226706	0.383797
22	0.534185	0.514142	0.884587	0.832513	0.760745	0.464523	0.742238
23	0.534219	0.528045	0.784282	0.582845	0.468901	0.279997	0.460106
24	0.533202	0.522678	0.915897	0.771349	0.711113	0.429696	0.715515
25	0.533930	0.498435	0.959848	0.624737	0.431162	0.238417	0.352147
26	0.533994	0.473308	0.751769	0.601362	0.497652	0.293719	0.466602
27	0.534347	0.490549	0.581880	0.399412	0.292939	0.187978	0.319654
28	0.534599	0.493209	0.577367	0.472474	0.381516	0.282428	0.398834
29	0.535016	0.542252	0.685416	0.583378	0.495894	0.301999	0.497886
..	...	...	...	...	...	...	...
569	0.961272	0.298420	0.784399	0.487087	0.295212	0.159479	0.233587
570	0.962754	0.292790	0.740670	0.625636	0.530771	0.316000	0.514563
571	0.962789	0.282405	0.640101	0.456091	0.306759	0.168550	0.256635
572	0.965795	0.273197	0.744007	0.610682	0.510313	0.308311	0.505444
573	0.965861	0.298286	0.945121	0.677498	0.501833	0.276979	0.418944
574	0.965932	0.289551	0.763559	0.571360	0.431723	0.249835	0.392768
575	0.967314	0.282025	0.908113	0.859399	0.812658	0.472620	0.750209
576	0.969025	0.297824	0.670475	0.420993	0.246618	0.132342	0.193857
577	0.969091	0.297849	0.972531	0.713005	0.566017	0.317615	0.498896
578	0.973402	0.287981	0.952672	0.610304	0.392882	0.210089	0.310350
579	0.973802	0.291776	0.983897	0.640654	0.421975	0.228130	0.338799
580	0.975476	0.309049	0.970518	0.768741	0.643327	0.370563	0.582424
581	0.978538	0.278035	0.645217	0.508073	0.402031	0.235801	0.388201
582	0.979157	0.284631	0.555309	0.348318	0.187399	0.099225	0.146726
583	0.980869	0.300957	0.919999	0.909828	0.807538	0.490651	0.816769
584	0.989807	0.285942	0.929251	0.717105	0.592950	0.351010	0.718175
585	0.994287	0.299594	0.807943	0.512747	0.340402	0.190015	0.280525
586	0.998599	0.282379	0.865822	0.596445	0.429176	0.237343	0.363350
587	0.568390	0.419475	0.606672	0.474869	0.487552	0.327585	0.560494
588	0.568524	0.448689	0.861767	0.691489	0.595357	0.358644	0.587459
589	0.568561	0.432060	0.566724	0.475430	0.396477	0.248257	0.419144
590	0.568668	0.454937	0.481789	0.401388	0.310106	0.187784	0.325742
591	0.569052	0.458284	0.266239	0.138565	0.000000	0.000000	0.000000
592	0.568872	0.427107	0.925533	0.736961	0.584714	0.332526	0.517608
593	0.826940	0.457255	0.914667	0.848939	0.805630	0.504494	0.843520
594	0.826690	0.425715	0.965916	0.734775	0.580787	0.325037	0.513312

595	0.252242	0.597662	0.917396	0.718748	0.576439	0.325313	0.505539
596	0.252272	0.608786	0.937480	0.742275	0.624845	0.370213	0.600172
597	0.254428	0.572213	0.891451	0.881503	0.776640	0.459835	0.734820
598	0.254739	0.607661	0.913808	0.696634	0.564298	0.332213	0.532963

	run	camcol	field	specobjid	class	redshift	plate	mjd	\
0	752.0	4	267	3.720000e+18	STAR	0.000305	3306	0.645394	
1	752.0	4	267	3.640000e+17	STAR	0.000288	323	0.000000	
2	752.0	4	268	3.230000e+17	GALAXY	0.043760	287	0.079625	
3	752.0	4	269	3.720000e+18	STAR	0.000269	3306	0.645394	
4	752.0	4	269	3.720000e+18	STAR	0.000516	3306	0.645394	
5	752.0	4	269	3.650000e+17	STAR	0.000419	324	0.009953	
6	752.0	4	269	3.230000e+17	GALAXY	0.035688	287	0.079625	
7	752.0	4	269	3.720000e+18	STAR	0.000419	3306	0.645394	
8	752.0	4	270	3.640000e+17	STAR	0.000339	323	0.000000	
9	752.0	4	270	3.240000e+17	GALAXY	0.014605	288	0.075137	
10	752.0	4	270	3.720000e+18	STAR	0.000296	3306	0.645394	
11	752.0	4	271	3.720000e+18	STAR	0.000528	3306	0.645394	
12	752.0	4	271	3.720000e+18	STAR	0.000327	3306	0.645394	
13	752.0	4	271	3.720000e+18	STAR	0.000311	3306	0.645394	
14	752.0	4	271	3.240000e+17	GALAXY	0.025751	288	0.075137	
15	752.0	4	272	3.230000e+17	QSO	0.096288	287	0.079625	
16	752.0	4	272	3.650000e+17	STAR	0.000382	324	0.009953	
17	752.0	4	272	3.230000e+17	QSO	0.416117	287	0.079625	
18	752.0	4	273	3.640000e+17	STAR	0.000393	323	0.000000	
19	752.0	4	273	3.230000e+17	GALAXY	0.025993	287	0.079625	
20	752.0	4	273	3.650000e+17	STAR	0.000212	324	0.009953	
21	752.0	4	274	3.240000e+17	STAR	0.000295	288	0.075137	
22	752.0	4	274	3.240000e+17	QSO	0.326847	288	0.075137	
23	752.0	4	274	2.880000e+18	STAR	0.000309	2558	0.492779	
24	752.0	4	274	3.720000e+18	STAR	0.000339	3306	0.645394	
25	752.0	4	274	3.240000e+17	GALAXY	0.041526	288	0.075137	
26	752.0	4	274	3.230000e+17	GALAXY	0.007824	287	0.079625	
27	752.0	4	275	2.880000e+18	STAR	0.000370	2558	0.492779	
28	752.0	4	275	3.720000e+18	STAR	0.000369	3306	0.645394	
29	752.0	4	275	2.880000e+18	STAR	0.000491	2558	0.492779	
..	...	...	...	...	...	...	...	...	
569	745.0	2	571	3.900000e+17	GALAXY	0.033064	346	0.015222	
570	745.0	2	573	3.900000e+17	GALAXY	0.006169	346	0.015222	
571	745.0	2	573	3.900000e+17	GALAXY	0.018124	346	0.015222	
572	745.0	2	576	3.900000e+17	STAR	0.000188	346	0.015222	
573	745.0	2	576	3.900000e+17	GALAXY	0.028194	346	0.015222	
574	745.0	2	576	3.900000e+17	GALAXY	0.021262	346	0.015222	
575	745.0	2	577	4.100000e+17	QSO	0.190299	364	0.075137	
576	745.0	2	579	4.100000e+17	GALAXY	0.021164	364	0.075137	
577	745.0	2	579	3.900000e+17	GALAXY	0.024719	346	0.015222	
578	745.0	2	583	4.100000e+17	GALAXY	0.033354	364	0.075137	
579	745.0	2	583	4.100000e+17	GALAXY	0.050581	364	0.075137	

580	745.0	2	585	4.100000e+17	GALAXY	0.016733	364	0.075137
581	745.0	2	588	4.100000e+17	GALAXY	0.016012	364	0.075137
582	745.0	2	588	4.100000e+17	GALAXY	0.016611	364	0.075137
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587	745.0	3	205	3.260000e+18	STAR	0.000537	2895	0.576112
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0.50449365, 0.50468837, 0.5099246 , 0.51119026, 0.51841408,
0.57123018, 0.57637549, 0.62184469, 1.      ])),
(0.10540330888896787, 0.282428095613796, 0.9865696042348661))

```

```

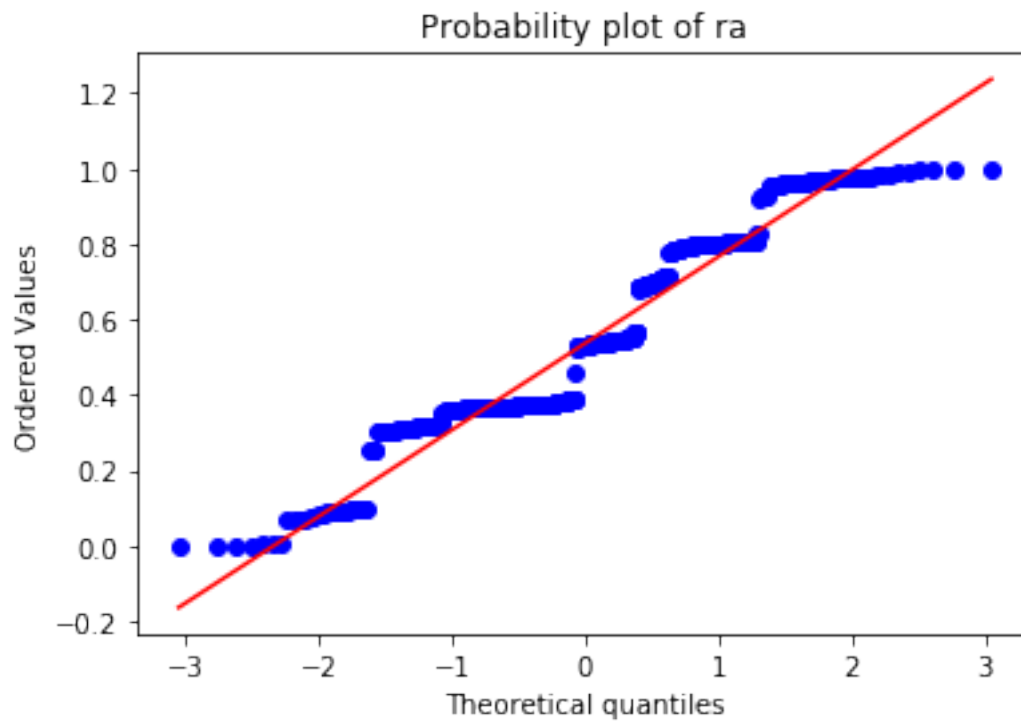
In [30]: res = scipy.stats.probplot(df['ra'], plot=plt)
plt.title("Probability plot of ra")

```

```

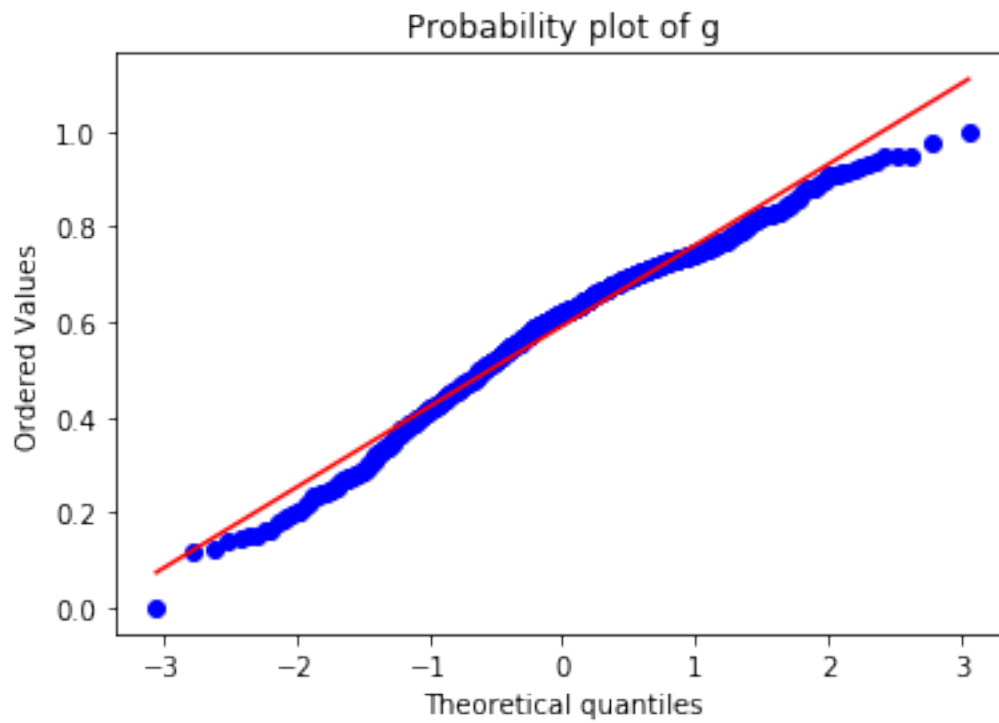
Out[30]: Text(0.5,1,'Probability plot of ra')

```



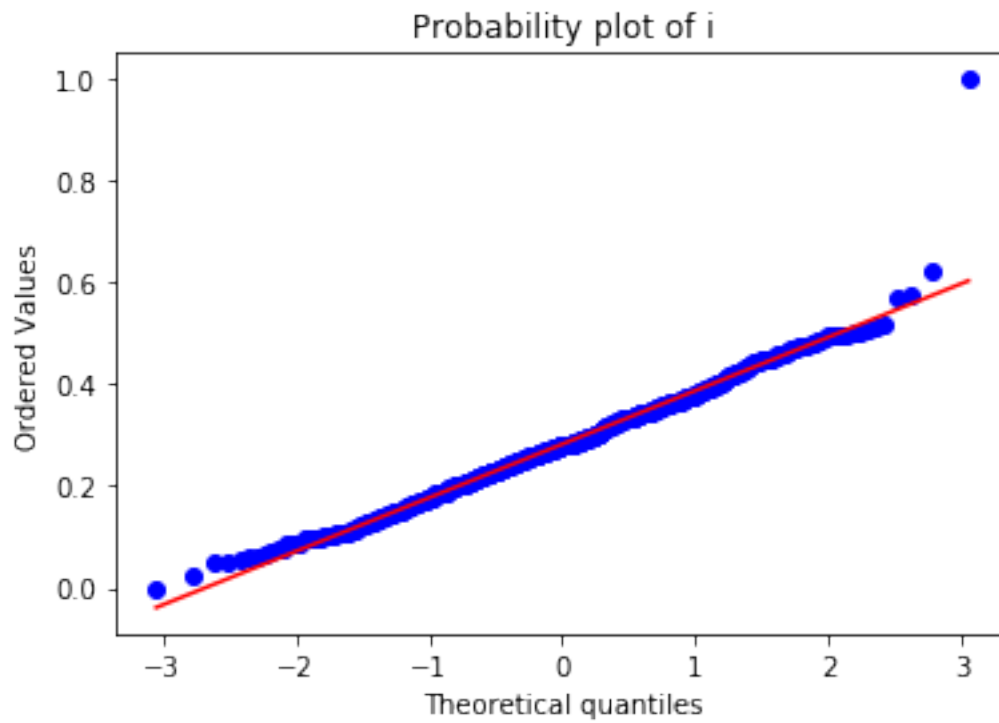
```
In [31]: res = scipy.stats.probplot(df['g'], plot=plt)
         plt.title("Probability plot of g")
```

```
Out[31]: Text(0.5,1,'Probability plot of g')
```



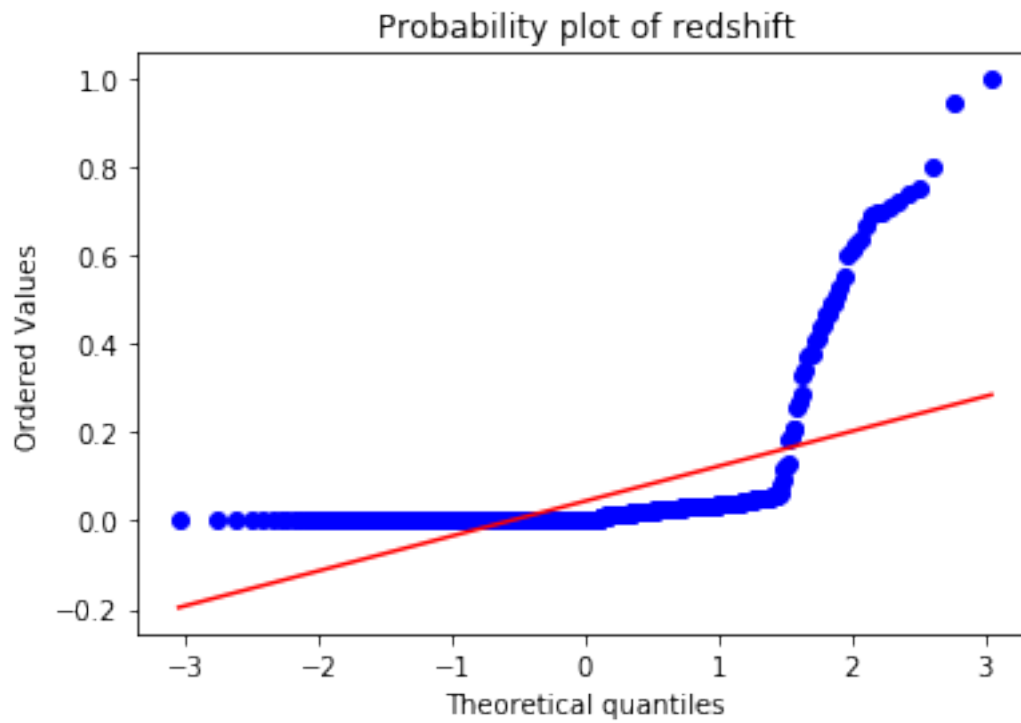
```
In [32]: res = scipy.stats.probplot(df['i'], plot=plt)
         plt.title("Probability plot of i")
```

```
Out[32]: Text(0.5,1,'Probability plot of i')
```



```
In [33]: res = scipy.stats.probplot(df['redshift'], plot=plt)
         plt.title("Probability plot of redshift")
```

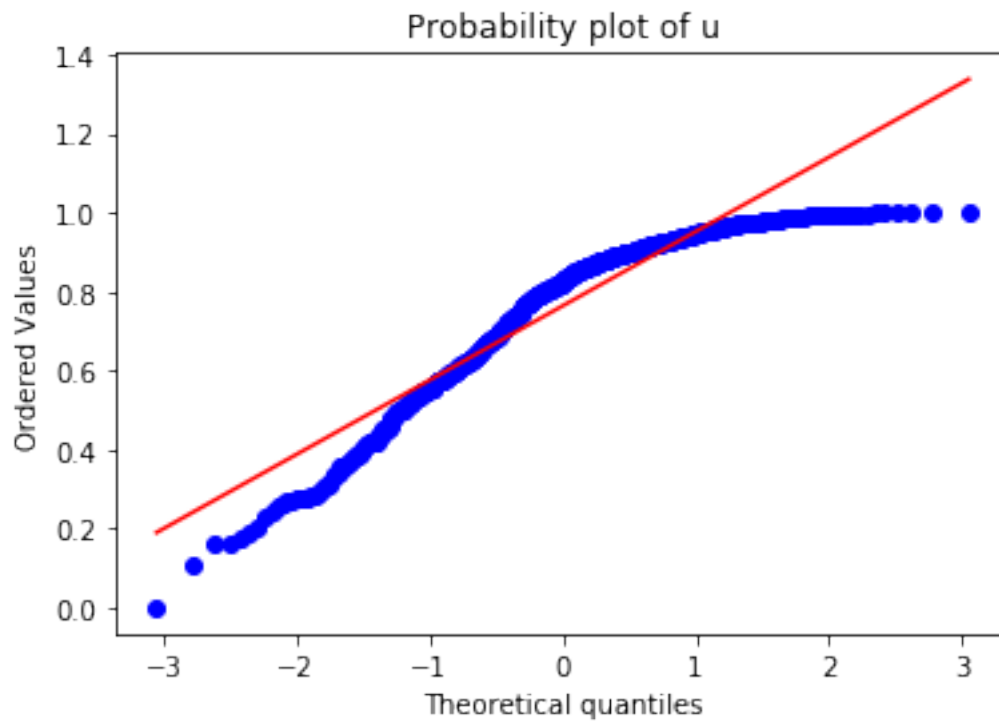
```
Out[33]: Text(0.5,1,'Probability plot of redshift')
```



```
In [34]: res = scipy.stats.probplot(df['u'], plot=plt)
         plt.title("Probability plot of u")
```

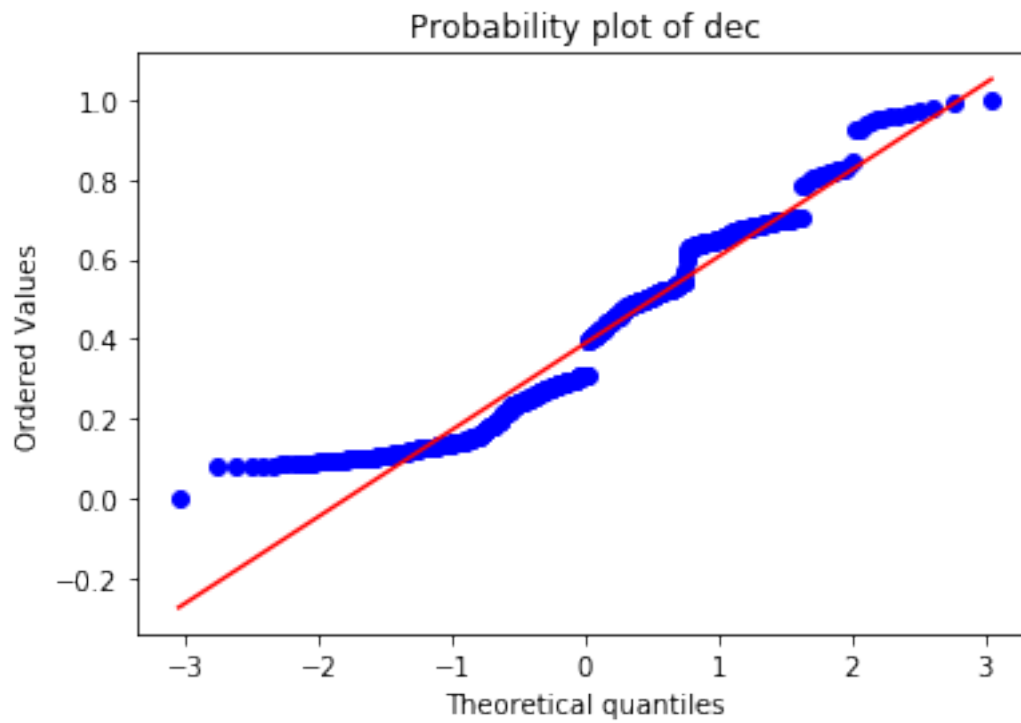
```
Out[34]: Text(0.5,1,'Probability plot of u')
```





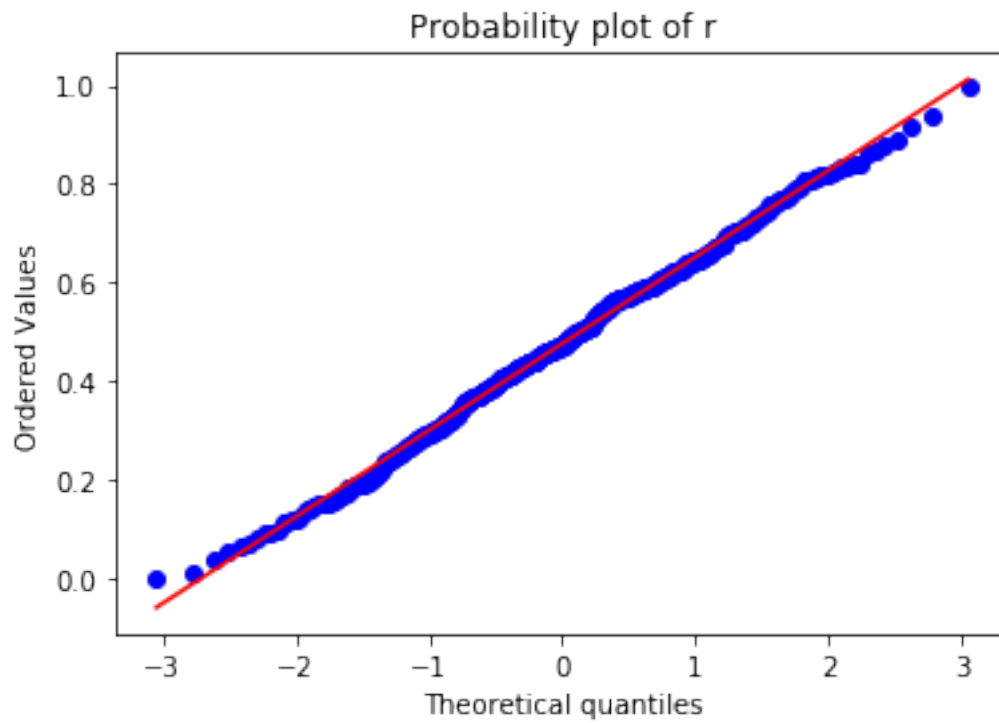
```
In [35]: res = scipy.stats.probplot(df['dec'], plot=plt)
         plt.title("Probability plot of dec")
```

```
Out[35]: Text(0.5,1,'Probability plot of dec')
```



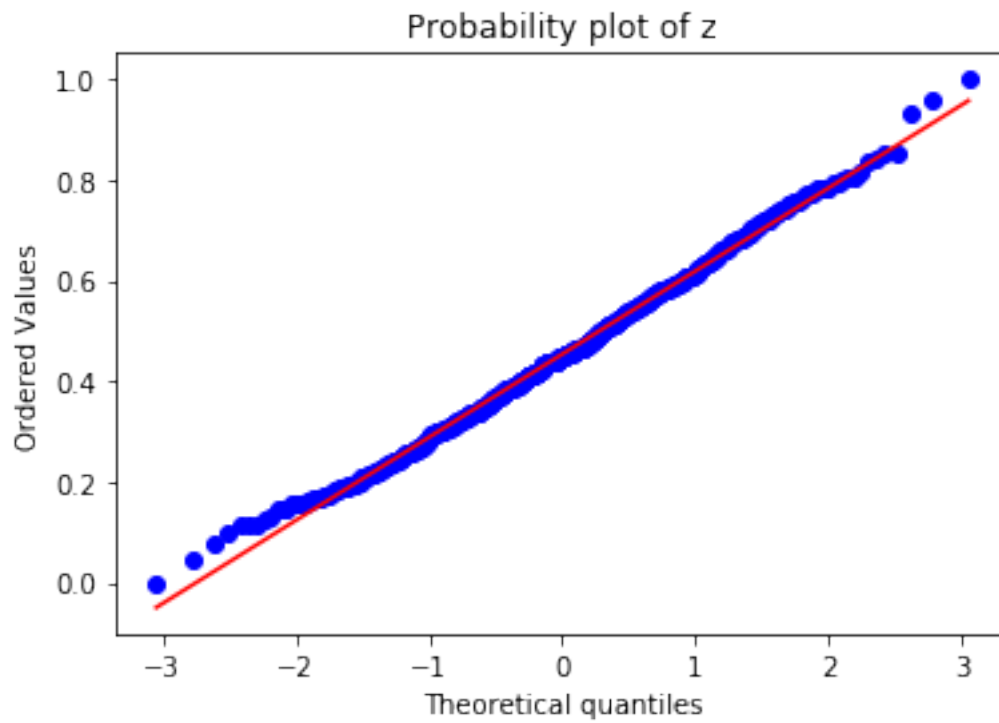
```
In [36]: res = scipy.stats.probplot(df['r'], plot=plt)
         plt.title("Probability plot of r")
```

```
Out[36]: Text(0.5,1,'Probability plot of r')
```



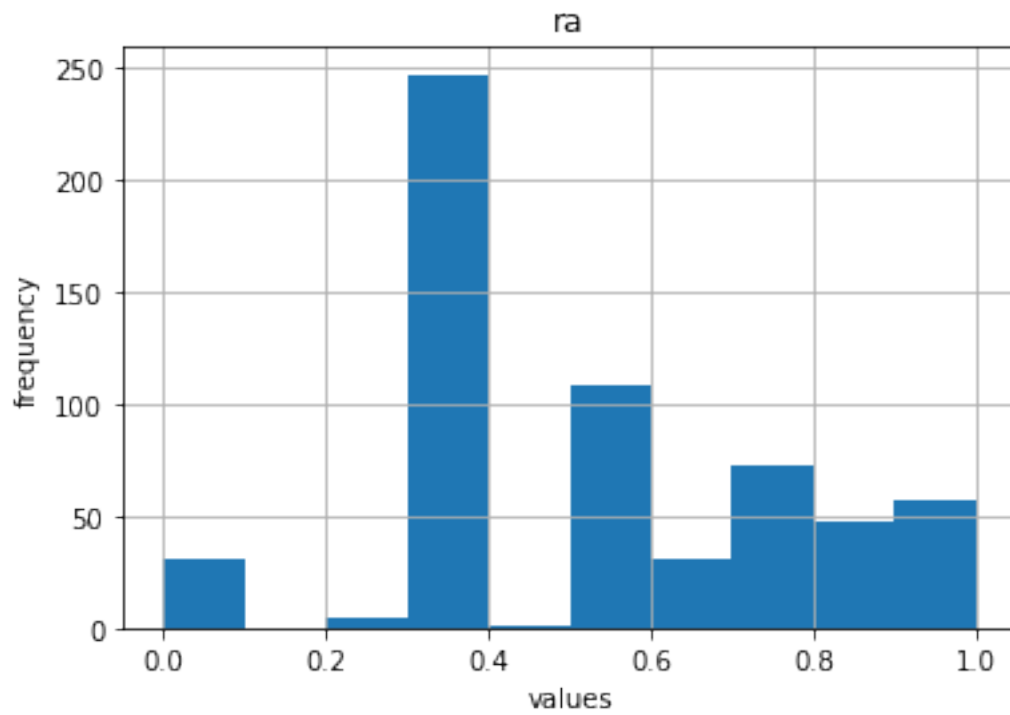
```
In [37]: res = scipy.stats.probplot(df['z'], plot=plt)
         plt.title("Probability plot of z")
```

```
Out[37]: Text(0.5,1,'Probability plot of z')
```



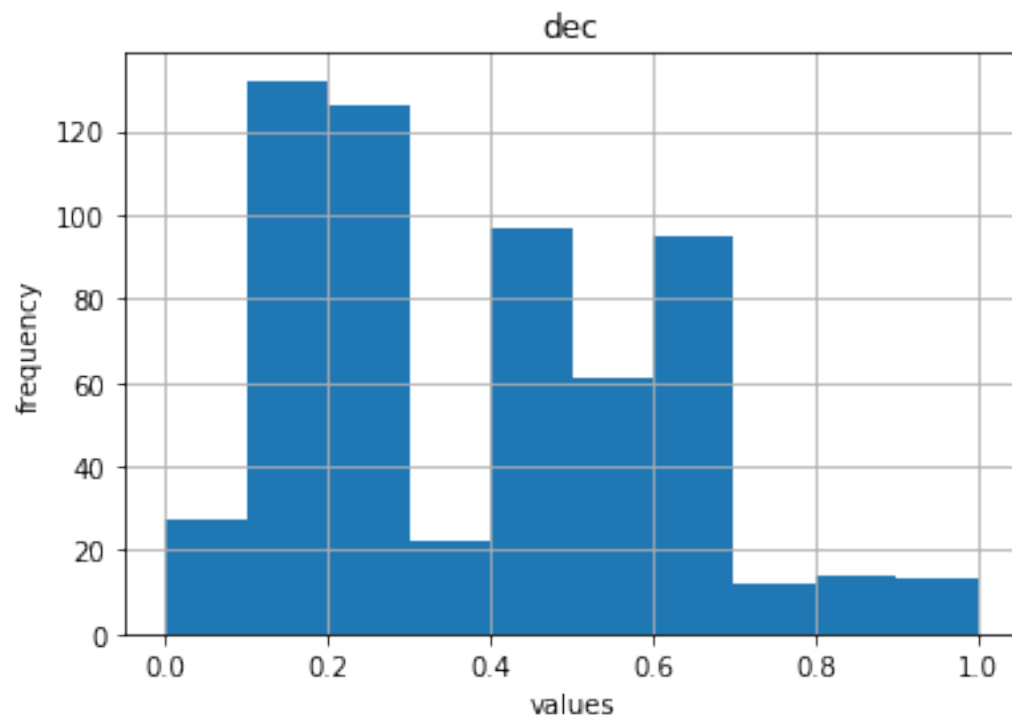
```
In [38]: df.hist(column='ra')  
         pl.ylabel("frequency")  
         pl.xlabel("values")
```

```
Out[38]: Text(0.5,0,'values')
```



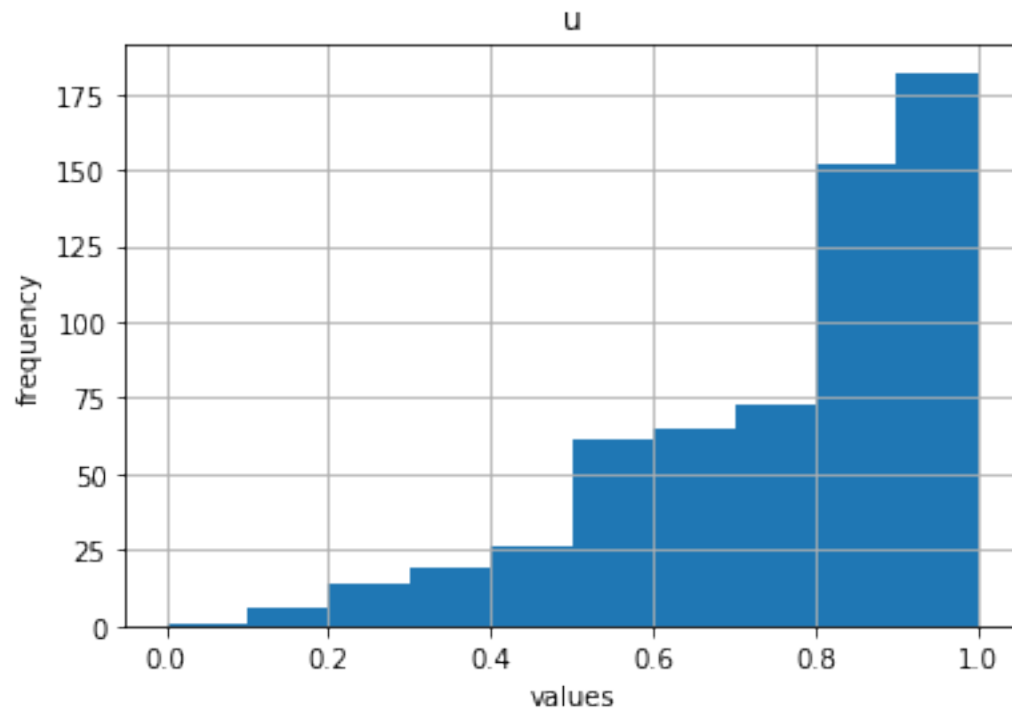
```
In [39]: df.hist(column='dec')  
         pl.ylabel("frequency")  
         pl.xlabel("values")
```

```
Out[39]: Text(0.5,0,'values')
```



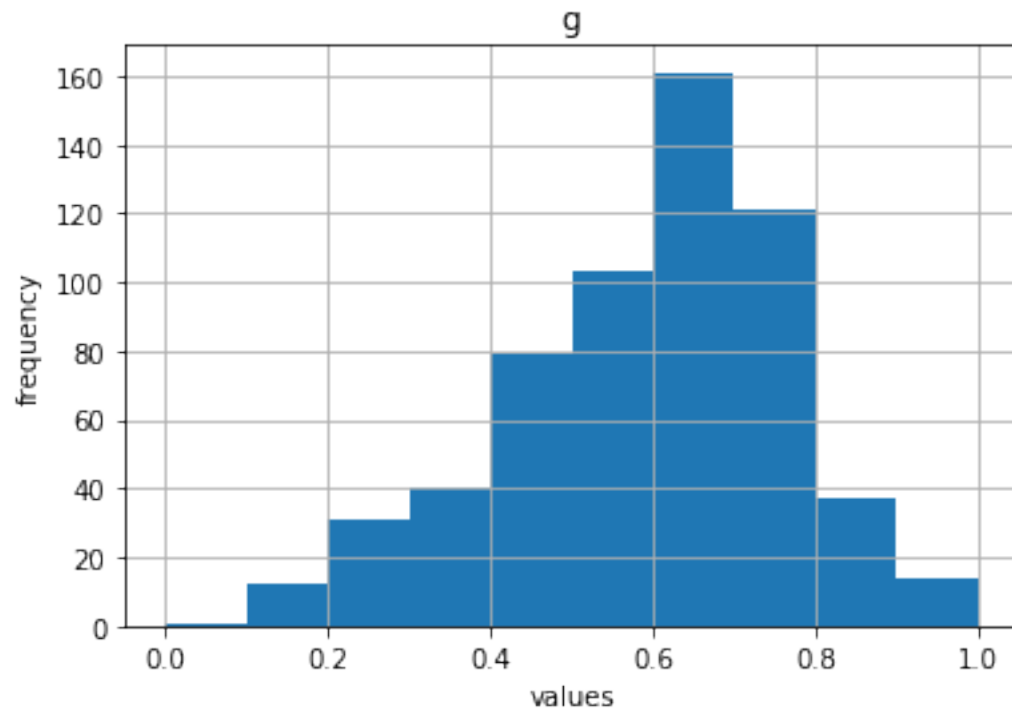
```
In [40]: df.hist(column='u')  
         pl.ylabel("frequency")  
         pl.xlabel("values")
```

```
Out[40]: Text(0.5,0,'values')
```



```
In [41]: df.hist(column='g')  
         pl.ylabel("frequency")  
         pl.xlabel("values")
```

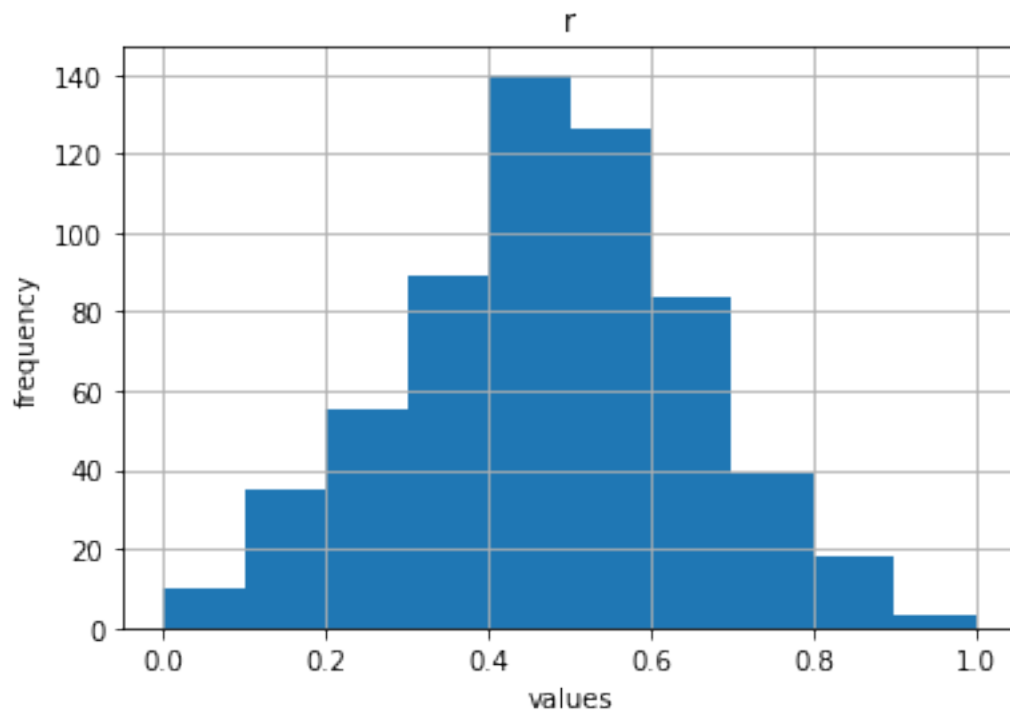
```
Out[41]: Text(0.5,0,'values')
```



```
In [42]: df.hist(column='r')  
         pl.ylabel("frequency")  
         pl.xlabel("values")
```

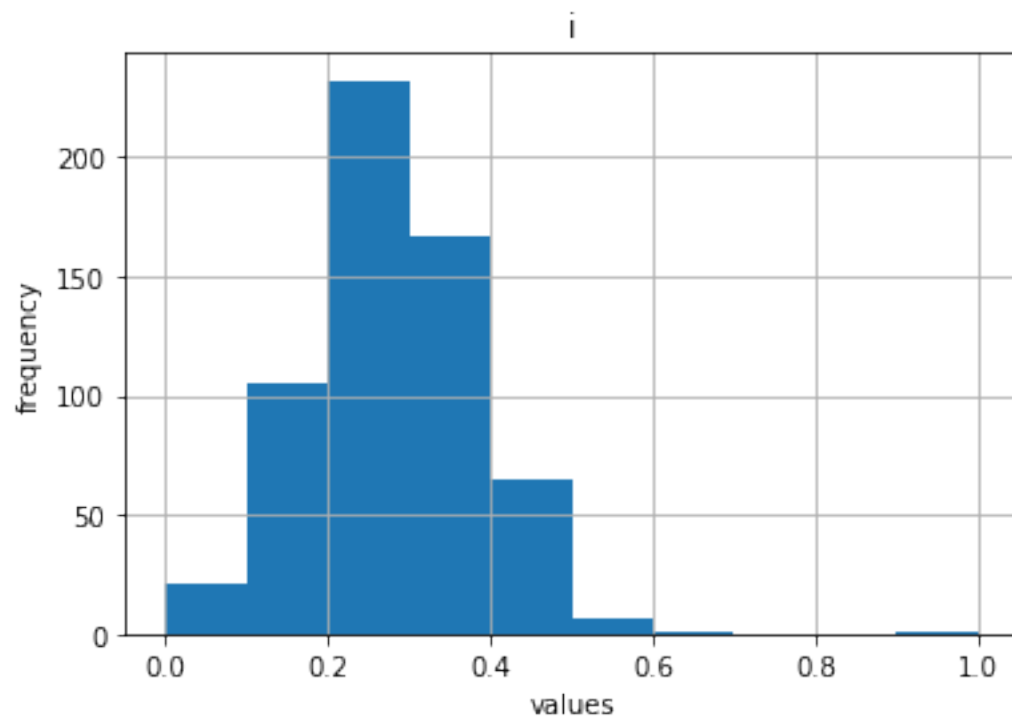
```
Out[42]: Text(0.5,0,'values')
```





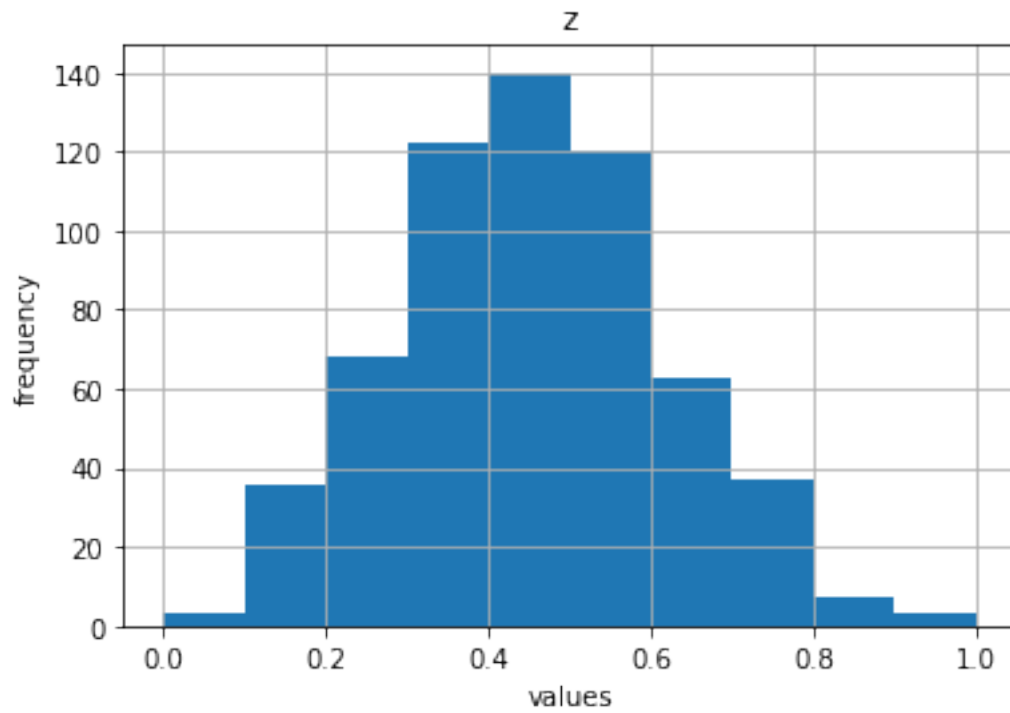
```
In [43]: df.hist(column='i')  
         pl.ylabel("frequency")  
         pl.xlabel("values")
```

```
Out[43]: Text(0.5,0,'values')
```



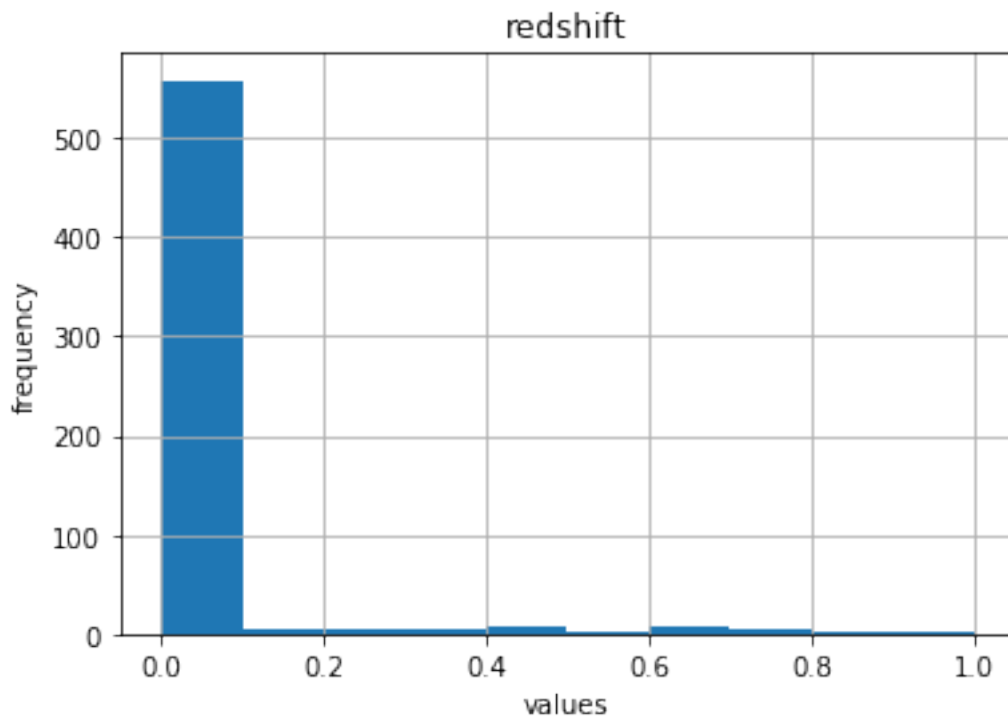
```
In [44]: df.hist(column='z')  
         pl.ylabel("frequency")  
         pl.xlabel("values")
```

```
Out[44]: Text(0.5,0,'values')
```



```
In [45]: df.hist(column='redshift')  
         pl.ylabel("frequency")  
         pl.xlabel("values")
```

```
Out[45]: Text(0.5,0,'values')
```



```
In [46]: # 3) Graph visualization
```

```
In [37]: #loading cleaned dataset
df1 = pd.read_csv('cleanedsky.csv')
```

```
In [48]: df1
```

```
Out[48]:
```

	ra	dec	u	g	r	i	z \
0	183.531326	0.089693	19.47406	17.04240	15.94699	15.503420	15.22531
1	183.598370	0.135285	18.66280	17.21449	16.67637	16.544191	16.39150
2	183.680207	0.126185	19.38298	18.19169	17.47428	17.087320	16.80125
3	183.870529	0.049911	17.76536	16.60272	16.16116	15.982330	15.90438
4	183.883288	0.102557	17.55025	16.26342	16.43869	16.544191	16.61326
5	183.847174	0.173694	19.43133	18.46779	18.16451	18.014750	18.04155
6	183.864379	0.019201	19.38322	17.88995	17.10537	16.663930	16.36955
7	183.900081	0.187473	18.97993	17.84496	17.38022	17.206730	17.07071
8	183.924588	0.097246	17.90616	16.97172	16.67541	16.544191	16.47596
9	183.973498	0.081626	18.67249	17.71375	17.49362	17.282840	17.22644
10	183.979195	0.135998	19.29772	17.80227	17.18266	16.923350	16.79928
11	184.085331	0.112110	18.83307	17.51785	16.94273	16.714180	16.60521
12	184.102098	0.191511	19.56250	18.19113	17.65759	17.475730	17.39203
13	184.160510	0.075645	19.57990	17.72815	16.98740	16.680760	16.50426
14	184.189574	0.099482	19.25667	17.54869	16.63578	16.149220	15.76639
15	184.350647	0.207230	18.73832	18.60962	18.39696	18.311740	17.97663

16	184.221797	0.046955	19.05958	18.09512	17.92766	16.544191	17.90772
17	184.245664	0.198257	19.22143	19.30248	19.13823	19.113510	19.23454
18	184.388870	0.068287	19.04397	17.51106	16.87335	16.611140	16.48303
19	184.380919	0.174323	17.81661	16.86976	16.53884	16.195760	16.08668
20	184.466853	0.111965	19.39320	18.48274	18.16551	18.051220	18.04328
21	184.569411	0.137091	17.51339	16.41793	16.06695	15.937510	15.89478
22	184.654170	0.122673	19.07731	18.64518	18.49678	18.526770	18.45765
23	184.658795	0.159936	18.63032	17.23437	16.72749	16.517720	16.44039
24	184.516834	0.145550	19.21684	18.29956	18.19589	18.147580	18.26658
25	184.618563	0.080576	19.41270	17.47109	16.49870	16.065020	15.66848
26	184.627408	0.013234	18.48543	17.33900	16.90179	16.667120	16.48684
27	184.676762	0.059442	17.72835	16.19783	15.66073	15.515860	15.43615
28	184.711970	0.066572	17.70824	16.61069	16.19772	16.544191	16.00229
29	184.770204	0.198010	18.18974	17.23738	16.89113	16.757270	16.71052
..	...	...	...	...	...	...	...
569	244.284910	-0.455477	18.63084	16.69326	15.67451	15.205570	14.82077
570	244.491831	-0.470566	18.43597	17.47617	17.10257	16.909710	16.82976
571	244.496791	-0.498398	17.98780	16.51811	15.74451	15.304330	14.98556
572	244.916397	-0.523076	18.45084	17.39167	16.97855	16.825990	16.76456
573	244.925582	-0.455836	19.34707	17.76923	16.92714	16.484860	16.14608
574	244.935545	-0.479247	18.53797	17.16947	16.50210	16.189330	15.95892
575	245.128493	-0.499415	19.18215	18.79711	18.81150	18.614920	18.51464
576	245.367353	-0.457074	18.12316	16.31978	15.37991	14.910120	14.53670
577	245.376694	-0.457008	19.46922	17.96987	17.31625	16.927290	16.71774
578	245.978479	-0.483455	19.38072	17.38953	16.26663	15.756590	15.36963
579	246.034378	-0.473284	19.51987	17.56103	16.44300	15.953020	15.57304
580	246.268157	-0.426990	19.46025	18.28482	17.78494	17.503770	17.31497
581	246.695626	-0.510111	18.01060	16.81185	16.32209	16.036530	15.92627
582	246.782081	-0.492432	17.60994	15.90911	15.02090	14.549550	14.19971
583	247.021139	-0.448677	19.23512	19.08207	18.78046	18.811240	18.99055
584	248.269016	-0.488920	19.27635	17.99304	17.47953	17.290880	18.28560
585	248.894520	-0.452329	18.73576	16.83826	15.94847	15.538030	15.15638
586	249.496551	-0.498468	18.99369	17.31122	16.48666	16.053320	15.74858
587	189.429821	-0.131042	17.83883	16.62422	16.84056	17.035840	17.15817
588	189.448544	-0.052745	18.97562	17.84829	17.49412	17.374000	17.35097
589	189.453801	-0.097313	17.66081	16.62739	16.28842	16.172150	16.14751
590	189.468747	-0.036000	17.28231	16.20900	15.76480	15.513750	15.47968
591	189.522249	-0.027031	16.32175	14.72385	13.88480	13.469230	13.15061
592	189.497150	-0.110585	19.25978	18.10524	17.42960	17.089640	16.85153
593	225.529253	-0.029788	19.21136	18.73800	18.76889	18.961950	19.18182
594	225.494356	-0.114317	19.43974	18.09289	17.40579	17.008100	16.82082
595	145.288576	0.346512	19.22352	18.00232	17.37943	17.011100	16.76524
596	145.292821	0.376326	19.31302	18.13527	17.67289	17.499960	17.44187
597	145.593807	0.278308	19.10790	18.92201	18.59314	18.475730	18.40461
598	145.637275	0.373311	19.20753	17.87736	17.30583	17.086230	16.96132

	run	camcol	field	specobjid	class	redshift	plate	mjd \
0	752.0	4	267	3.720000e+18	STAR	-8.960000e-06	3306	54922

1	752.0	4	267	3.640000e+17	STAR	-5.490000e-05	323	51615
2	752.0	4	268	3.230000e+17	GALAXY	1.231112e-01	287	52023
3	752.0	4	269	3.720000e+18	STAR	-1.106160e-04	3306	54922
4	752.0	4	269	3.720000e+18	STAR	5.903570e-04	3306	54922
5	752.0	4	269	3.650000e+17	STAR	3.146030e-04	324	51666
6	752.0	4	269	3.230000e+17	GALAXY	1.002423e-01	287	52023
7	752.0	4	269	3.720000e+18	STAR	3.148480e-04	3306	54922
8	752.0	4	270	3.640000e+17	STAR	8.910000e-05	323	51615
9	752.0	4	270	3.240000e+17	GALAXY	4.050813e-02	288	52000
10	752.0	4	270	3.720000e+18	STAR	-3.460000e-05	3306	54922
11	752.0	4	271	3.720000e+18	STAR	6.227740e-04	3306	54922
12	752.0	4	271	3.720000e+18	STAR	5.470000e-05	3306	54922
13	752.0	4	271	3.720000e+18	STAR	8.300000e-06	3306	54922
14	752.0	4	271	3.240000e+17	GALAXY	7.208736e-02	288	52000
15	752.0	4	272	3.230000e+17	QSO	2.719369e-01	287	52023
16	752.0	4	272	3.650000e+17	STAR	2.099740e-04	324	51666
17	752.0	4	272	3.230000e+17	QSO	1.178098e+00	287	52023
18	752.0	4	273	3.640000e+17	STAR	2.400980e-04	323	51615
19	752.0	4	273	3.230000e+17	GALAXY	7.277206e-02	287	52023
20	752.0	4	273	3.650000e+17	STAR	-2.702180e-04	324	51666
21	752.0	4	274	3.240000e+17	STAR	-3.550000e-05	288	52000
22	752.0	4	274	3.240000e+17	QSO	9.251733e-01	288	52000
23	752.0	4	274	2.880000e+18	STAR	4.310000e-06	2558	54140
24	752.0	4	274	3.720000e+18	STAR	8.800000e-05	3306	54922
25	752.0	4	274	3.240000e+17	GALAXY	1.167829e-01	288	52000
26	752.0	4	274	3.230000e+17	GALAXY	2.129653e-02	287	52023
27	752.0	4	275	2.880000e+18	STAR	1.753650e-04	2558	54140
28	752.0	4	275	3.720000e+18	STAR	1.723800e-04	3306	54922
29	752.0	4	275	2.880000e+18	STAR	5.198080e-04	2558	54140
..	...	...	...	...	...	...	...	...
569	745.0	2	571	3.900000e+17	GALAXY	9.280799e-02	346	51693
570	745.0	2	573	3.900000e+17	GALAXY	1.660742e-02	346	51693
571	745.0	2	573	3.900000e+17	GALAXY	5.047690e-02	346	51693
572	745.0	2	576	3.900000e+17	STAR	-3.407500e-04	346	51693
573	745.0	2	576	3.900000e+17	GALAXY	7.900908e-02	346	51693
574	745.0	2	576	3.900000e+17	GALAXY	5.936759e-02	346	51693
575	745.0	2	577	4.100000e+17	QSO	5.382954e-01	364	52000
576	745.0	2	579	4.100000e+17	GALAXY	5.909181e-02	364	52000
577	745.0	2	579	3.900000e+17	GALAXY	6.916405e-02	346	51693
578	745.0	2	583	4.100000e+17	GALAXY	9.362785e-02	364	52000
579	745.0	2	583	4.100000e+17	GALAXY	1.424379e-01	364	52000
580	745.0	2	585	4.100000e+17	GALAXY	4.653732e-02	364	52000
581	745.0	2	588	4.100000e+17	GALAXY	4.449500e-02	364	52000
582	745.0	2	588	4.100000e+17	GALAXY	4.619161e-02	364	52000
583	745.0	2	590	4.100000e+17	QSO	1.149604e+00	364	52000
584	745.0	2	598	3.920000e+17	STAR	-7.320000e-05	348	51671
585	745.0	2	602	3.920000e+17	GALAXY	7.018682e-02	348	51671
586	745.0	2	606	3.920000e+17	GALAXY	5.894065e-02	348	51671

587	745.0	3	205	3.260000e+18	STAR	6.484750e-04	2895	54567
588	745.0	3	205	3.280000e+17	STAR	2.573870e-04	291	51928
589	745.0	3	205	3.260000e+18	STAR	5.910000e-07	2895	54567
590	745.0	3	205	3.260000e+18	STAR	5.791130e-04	2895	54567
591	745.0	3	205	3.270000e+17	GALAXY	1.251297e-02	290	51941
592	745.0	3	205	3.280000e+17	GALAXY	1.383053e-01	291	51928
593	745.0	3	446	4.520000e+18	STAR	-1.976670e-04	4016	55632
594	745.0	3	446	3.490000e+17	GALAXY	4.019753e-02	310	51990
595	756.0	4	197	3.000000e+17	GALAXY	6.460119e-02	266	51630
596	756.0	4	197	3.000000e+17	STAR	2.133720e-04	266	51630
597	756.0	4	199	3.000000e+17	QSO	1.404700e+00	266	51630
598	756.0	4	199	3.000000e+17	STAR	-3.480000e-05	266	51630

	fiberid
0	491
1	541
2	513
3	510
4	512
5	594
6	559
7	515
8	595
9	400
10	506
11	547
12	544
13	546
14	389
15	587
16	35
17	583
18	628
19	632
20	623
21	430
22	421
23	356
24	595
25	437
26	637
27	358
28	632
29	349
..	...
569	162
570	148
571	146

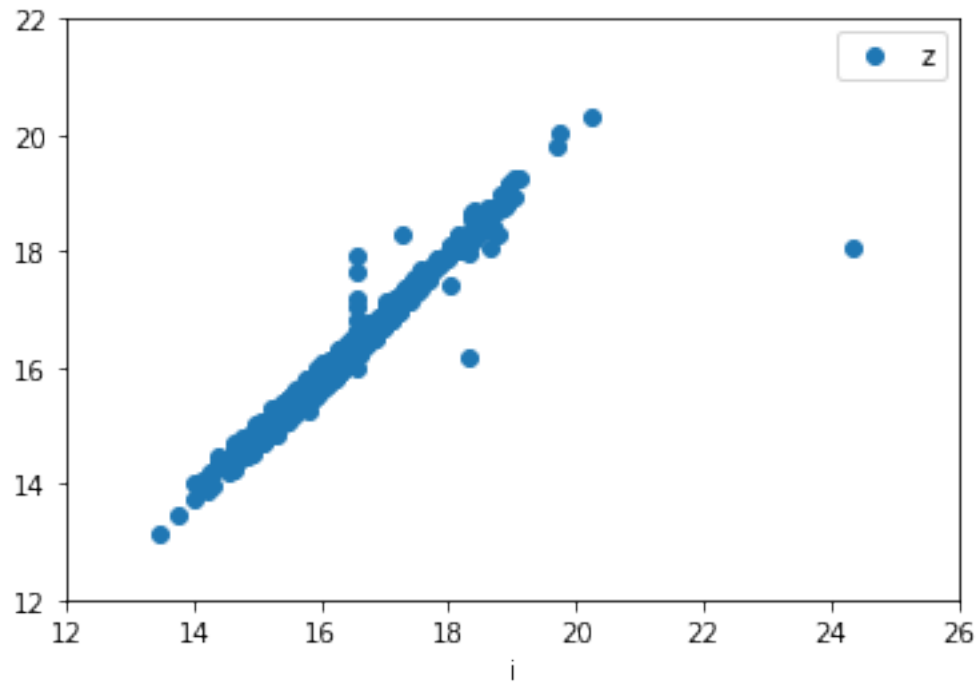
572	81
573	106
574	108
575	307
576	302
577	67
578	221
579	190
580	161
581	146
582	107
583	74
584	236
585	163
586	106
587	116
588	310
589	117
590	119
591	516
592	318
593	396
594	224
595	359
596	346
597	396
598	390

[599 rows x 16 columns]

```
In [49]: #Scatter Plot
df1.plot(x='i', y='z', style='o')
plt.xlim([12,26])
plt.ylim([12,22])
```

Out[49]: (12, 22)

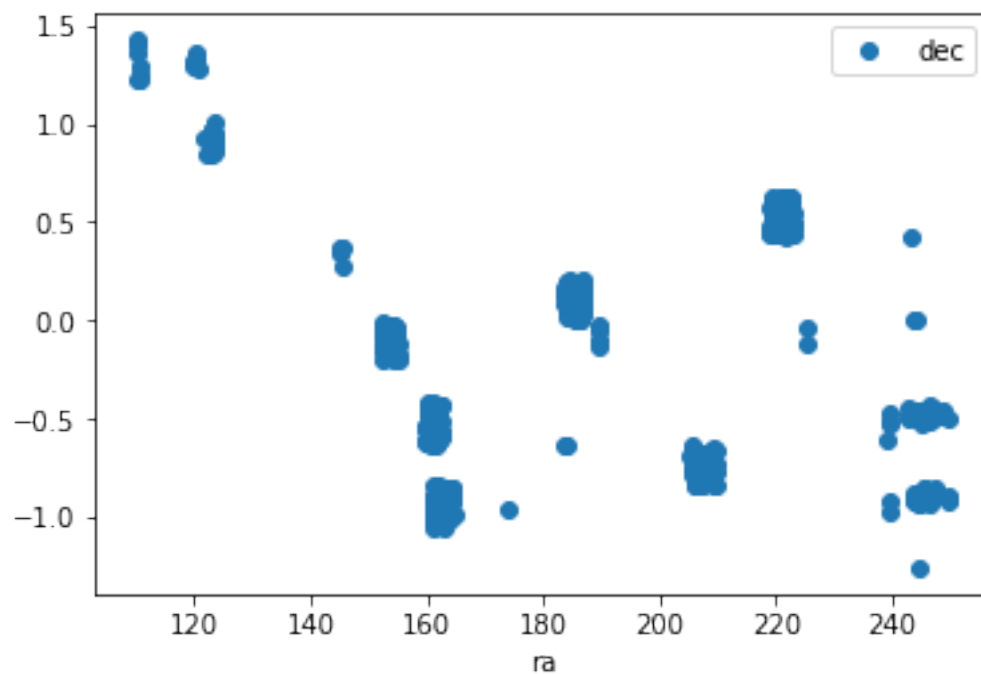




In [50]: *#We see that scatterplot for infrared and 900nm response of the telescope are correlated  
#This is clear as infrared wavelenghts(700nm and above) and 900nm are almost near hence*

In [51]: `df1.plot(x='ra', y='dec', style='o')`

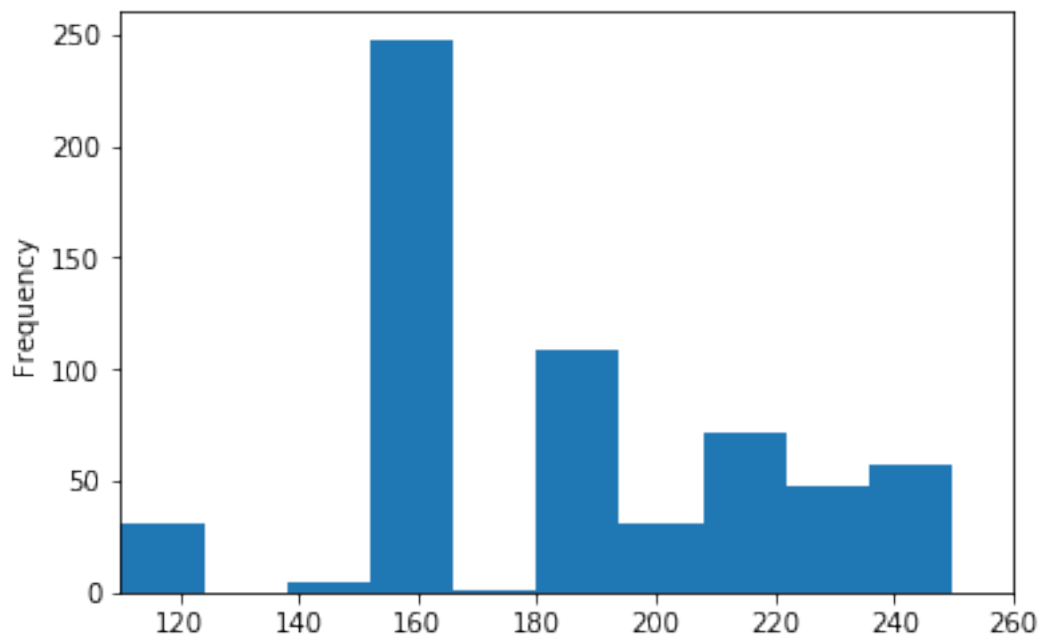
Out[51]: `<matplotlib.axes._subplots.AxesSubplot at 0x243aff6ca90>`



In [ ]: *#We see that ra and dec are not correlated this is obvious as the right ascension angle*

```
In [58]: #df1.hist(column='ra')
df1['ra'].plot.hist(xlim = (110,260),ylim=(0,260))
#plt.xlim([110,260])
#plt.ylim([0,260])
```

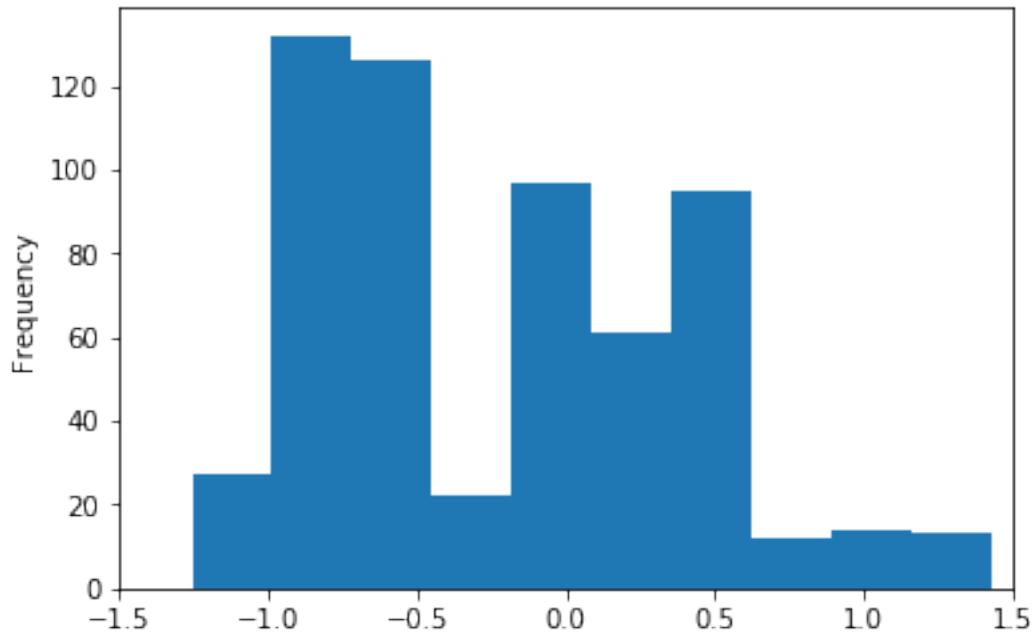
Out[58]: <matplotlib.axes.\_subplots.AxesSubplot at 0x243b040fda0>



In [118]: *# insights from this histogram plot shows that most of our data is taken from right as*

```
In [63]: #df.hist(column='dec')
df1['dec'].plot.hist(xlim=(-1.5,1.5))
```

Out[63]: <matplotlib.axes.\_subplots.AxesSubplot at 0x243b0923358>

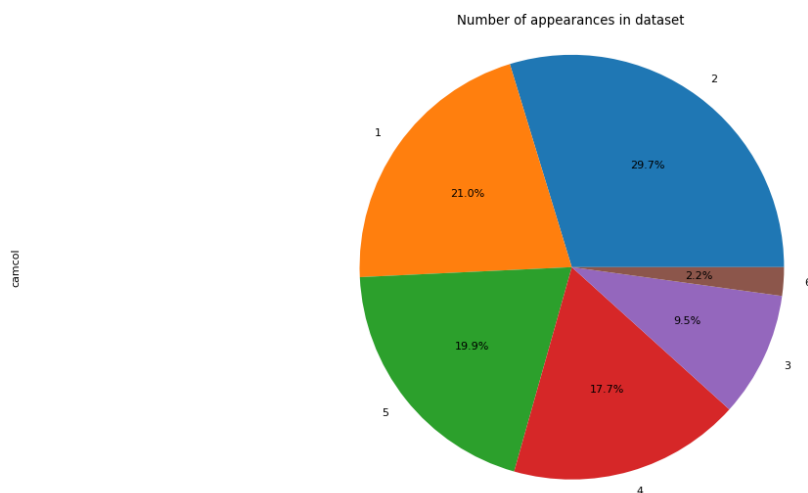


In [41]: # insights from this histogram plot shows that declination angle of our data is little

In [42]: # from both of this histograms of ra and dec we can clearly see that most of our dataset

```
In [40]: plt.figure(figsize=(18, 8), dpi=80, facecolor='w', edgecolor='k')
df1['camcol'].value_counts().plot(kind='pie', autopct='%1.1f%%')
plt.axis('equal')
plt.title('Number of appearances in dataset')

plt.show()
df1['camcol'].value_counts()
```



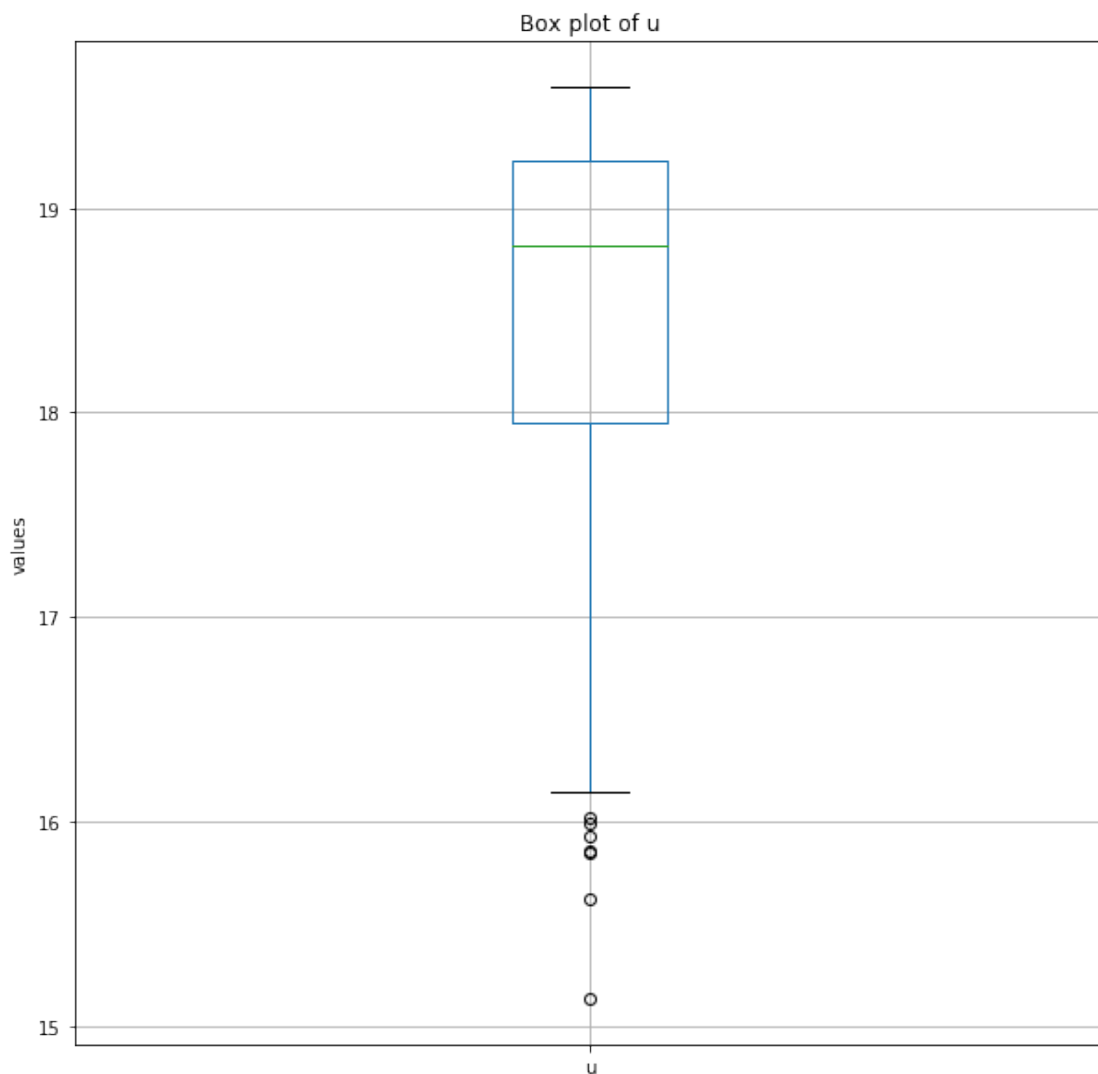
```
Out[40]: 2    178
         1    126
         5    119
         4    106
         3     57
         6     13
         Name: camcol, dtype: int64
```

```
In [44]: # We can see that most of our dataset contains Stars
```

```
In [45]: # very few 7% of our dataset contains quasars which says that quasars are very rare whe
```

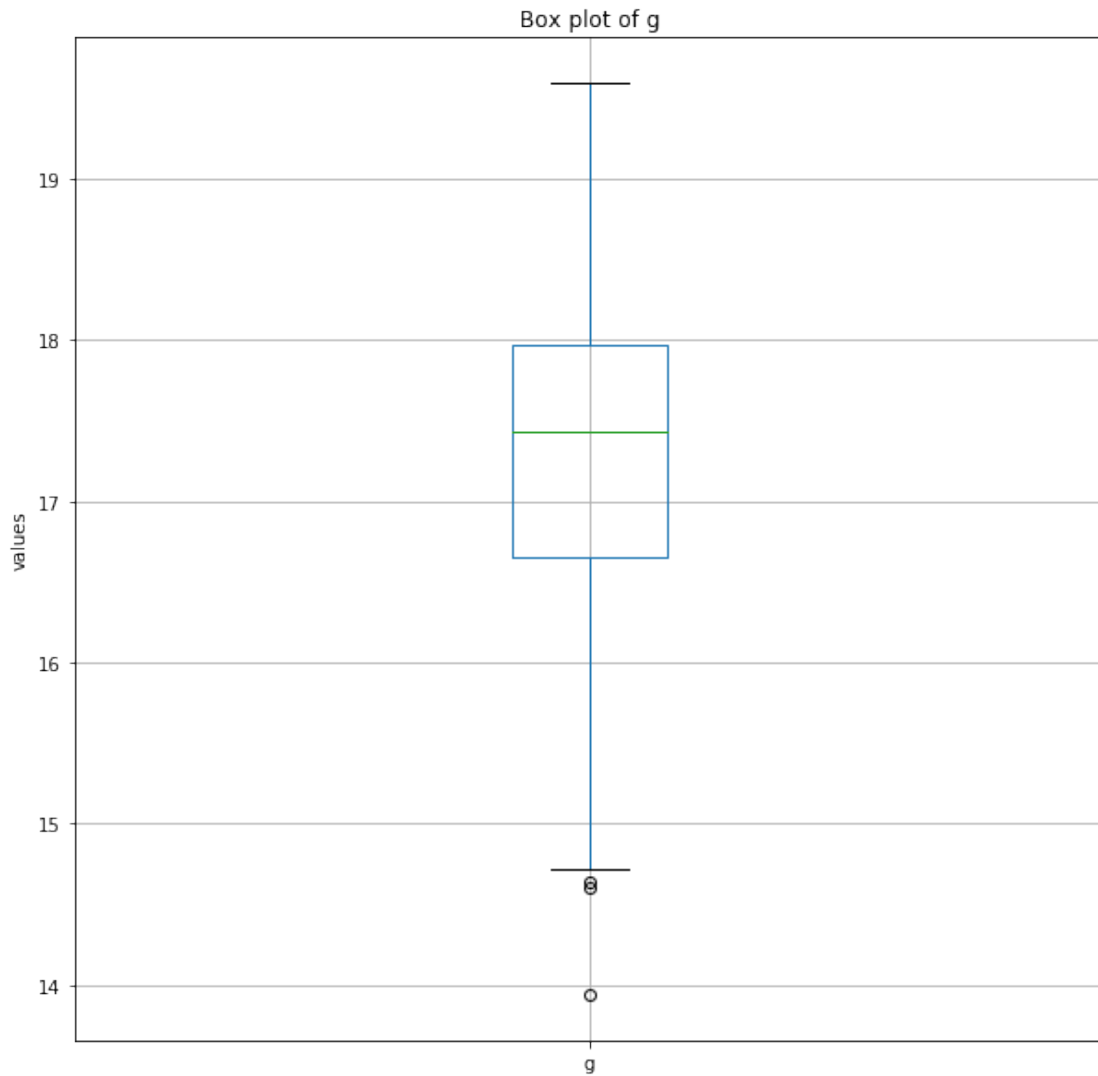
```
In [67]: df1.boxplot(column='u',figsize=(10,10))
         pl.title("Box plot of u")
         pl.ylabel("values")
```

```
Out[67]: Text(0,0.5,'values')
```



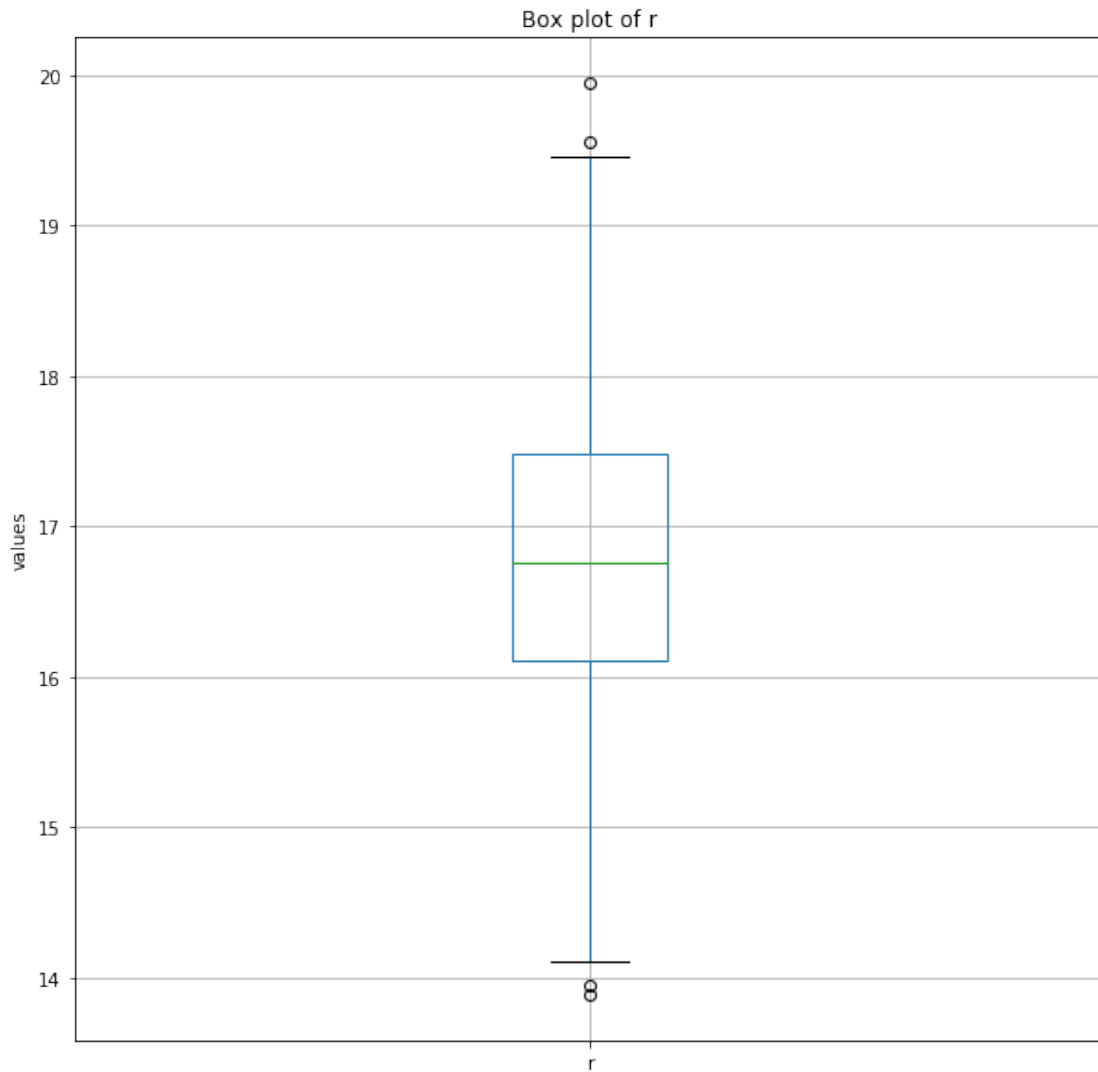
```
In [68]: #We can see that the response of the telescope to ultraviolet band ranges from 16.2 to
df1.boxplot(column='g',figsize=(10,10))
pl.title("Box plot of g")
pl.ylabel("values")

Out[68]: Text(0,0.5,'values')
```



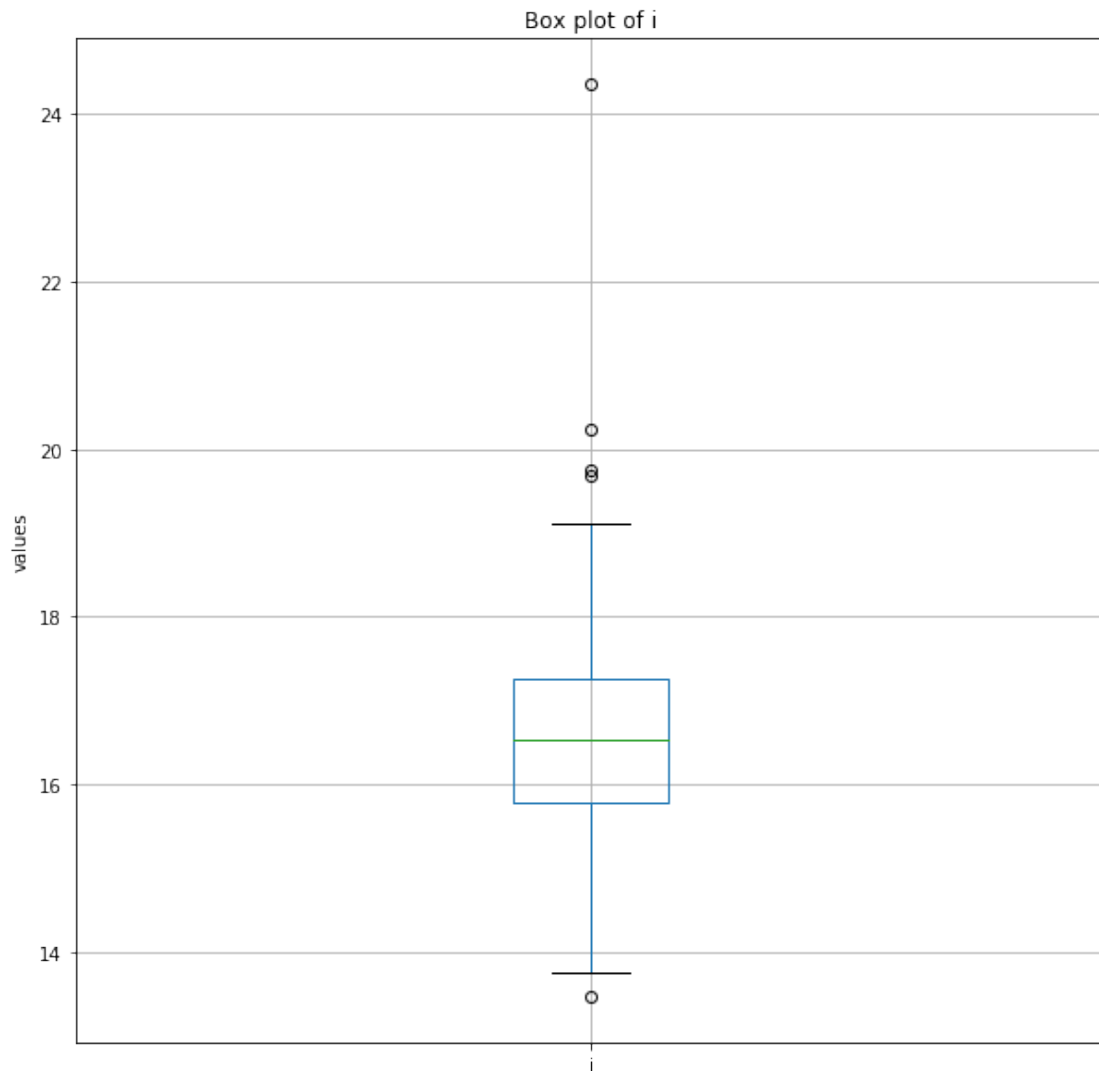
```
In [69]: #We can see that the response of the telescope to green band ranges from 14.7 to 19.6 u
df1.boxplot(column='r',figsize=(10,10))
pl.title("Box plot of r")
pl.ylabel("values")
```

Out[69]: Text(0,0.5,'values')



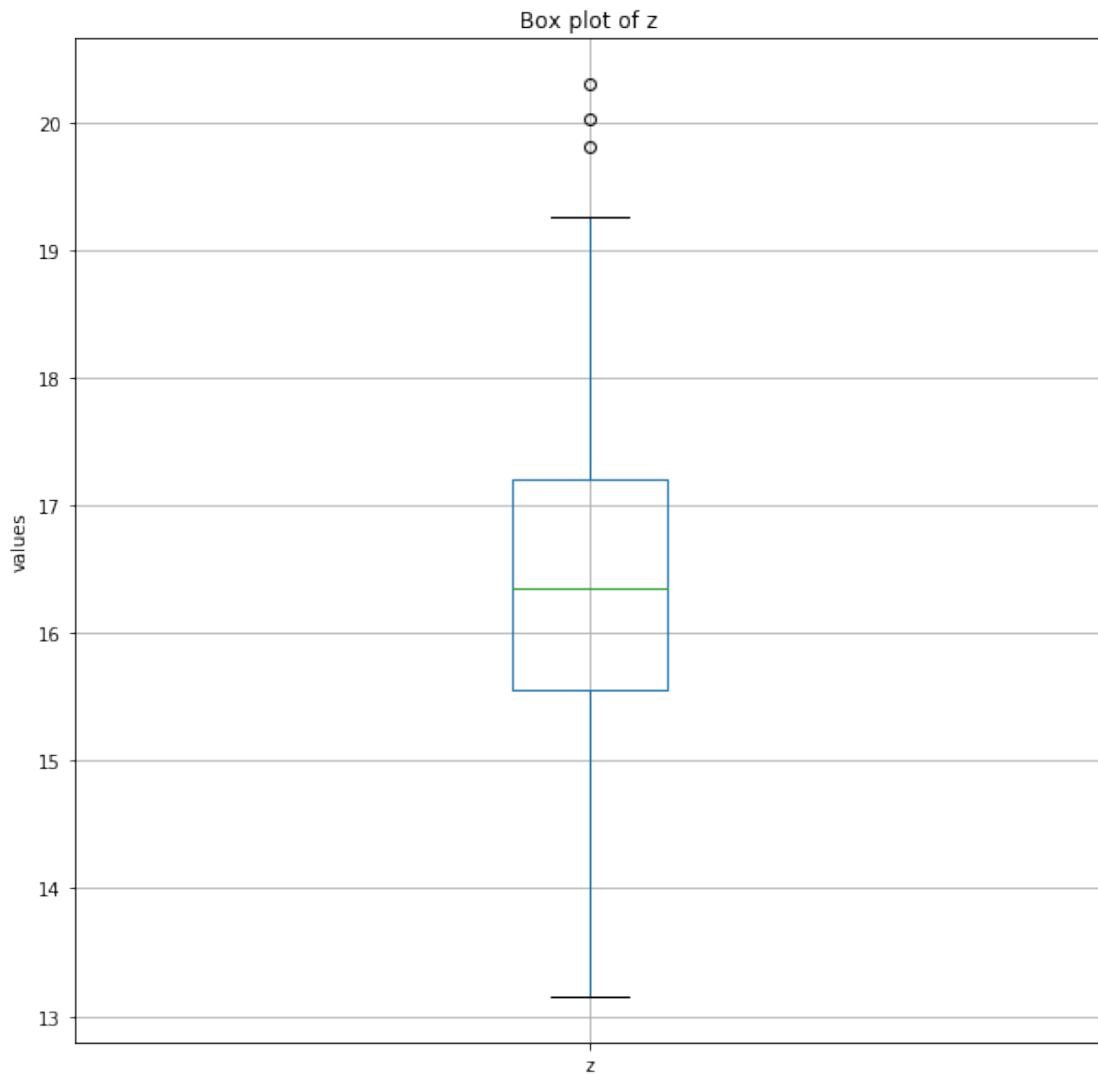
In [70]: *#We can see that the response of the telescope to red band ranges from 14.1 to 19.5 with*  
df1.boxplot(column='i',figsize=(10,10))  
pl.title("Box plot of i")  
pl.ylabel("values")

Out[70]: Text(0,0.5,'values')



```
In [71]: #We can see that the response of the telescope to infrared band ranges from 13.9 to 19
df1.boxplot(column='z',figsize=(10,10))
plt.title("Box plot of z")
plt.ylabel("values")
```

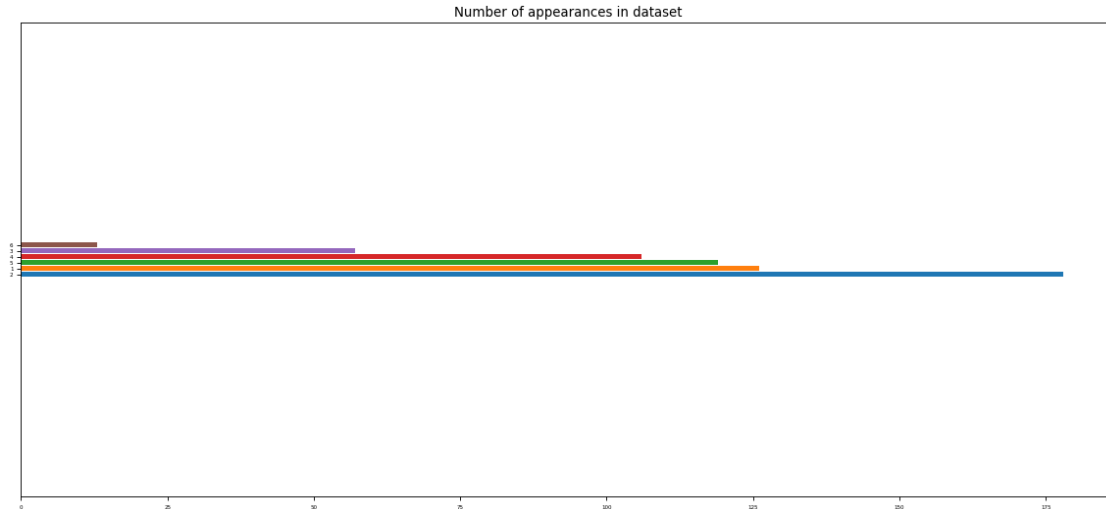
```
Out[71]: Text(0,0.5,'values')
```



In [51]: *#We can see that the response of the telescope to z(900nm) band ranges from 13.2 to 19.2*

```
In [76]: plt.figure(num=None, figsize=(18, 8), dpi=80, facecolor='w', edgecolor='k')
df1['camcol'].value_counts().plot(kind='barh', fontsize = 5, width = 0.8, rot=0)
plt.axis('equal')
plt.title('Number of appearances in dataset')
plt.show()
```





In [53]: *# We can see by this bar graph that camera scanline within in the run is used maximum f*

In [97]: *# 4) Hypothesis testing*

In [101]: *# picking a random sample of size 100 from our dataset with replacement as a specific*

In [99]: `sample = df1.sample(n = 100,replace = True)`

In [100]: `sample`

```
Out[100]:
```

	ra	dec	u	g	r	i	z	\
585	248.894520	-0.452329	18.73576	16.83826	15.94847	15.53803	15.15638	
89	160.996765	-0.588245	18.38625	16.39808	15.40492	14.88012	14.50048	
544	222.955420	0.472939	18.31584	17.36814	17.06714	16.91268	16.87696	
464	162.964354	-0.854905	15.92846	15.75834	15.68789	15.79105	15.25990	
105	161.303101	-0.964276	19.01558	17.72445	17.18152	16.97813	16.85652	
288	209.090702	-0.656281	19.23039	18.06944	17.59436	17.23812	17.09449	
182	221.163280	0.483958	17.88442	16.96119	16.64287	16.53269	16.48384	
472	163.162657	-0.943445	18.98361	17.83656	17.32091	17.11787	17.00929	
156	219.581971	0.471853	18.56313	18.53411	18.38498	18.37235	18.36230	
218	110.246454	1.364872	18.58476	17.29450	16.42206	16.27426	15.89856	
108	161.350692	-0.893119	18.12173	16.84951	16.38322	16.20272	16.10719	
447	162.670124	-0.874412	19.07583	17.58162	16.95899	16.70053	16.53613	
243	244.949053	-0.894235	19.15965	18.91703	18.61517	18.38544	18.42006	
51	159.797842	-0.557196	19.19752	18.07471	17.76889	17.61480	17.49577	
543	222.933968	0.432157	18.23652	17.03079	16.52341	16.28072	16.18811	
578	245.978479	-0.483455	19.38072	17.38953	16.26663	15.75659	15.36963	
239	243.847368	-0.898343	19.41207	18.88776	18.56437	18.58996	18.39260	
163	220.122411	0.584089	16.93221	15.56652	14.92086	14.58185	14.32520	
297	209.951317	-0.766163	19.15065	17.36285	16.52798	16.15387	15.80557	

178	221.041738	0.564546	18.77460	17.64604	17.22892	17.07624	17.01672	
162	219.938725	0.531203	18.91460	17.67455	16.96098	16.57482	16.29057	
532	222.509033	0.578781	17.78947	15.91691	15.07053	14.67859	14.36553	
25	184.618563	0.080576	19.41270	17.47109	16.49870	16.06502	15.66848	
414	153.449439	-0.174722	19.33492	17.25626	16.32616	15.91549	15.54986	
375	161.988413	-0.455833	15.84520	14.78328	14.48869	14.40722	14.38870	
29	184.770204	0.198010	18.18974	17.23738	16.89113	16.75727	16.71052	
267	206.806474	-0.671287	18.37380	16.84218	16.10014	15.71413	15.41793	
467	163.063223	-0.964999	17.85959	16.18301	15.56699	15.37634	15.28177	
234	174.164488	-0.956707	19.57255	18.59834	18.47911	18.45792	18.60459	
27	184.676762	0.059442	17.72835	16.19783	15.66073	15.51586	15.43615	
..	...	...	...	...	...	...	...	
553	247.254489	-0.854531	18.25633	16.75476	16.15203	15.92617	15.82208	
21	184.569411	0.137091	17.51339	16.41793	16.06695	15.93751	15.89478	
159	219.661166	0.569298	19.47595	18.61160	18.52926	18.76056	18.28581	
134	162.246566	-0.918542	17.71105	16.34725	15.83003	15.60209	15.47698	
158	219.588635	0.510879	19.14528	18.19967	17.89295	17.78337	17.73488	
499	221.766733	0.479062	17.68773	16.15192	15.37372	14.95320	14.64231	
188	221.364993	0.454015	17.52671	16.10266	15.57270	15.35048	15.24569	
222	110.469850	1.228435	19.42571	17.78678	17.11364	16.85967	16.67854	
399	152.289487	-0.063071	19.45274	17.62178	16.68552	16.27155	15.92060	
53	159.847524	-0.534266	19.38251	17.92302	17.11018	16.66492	16.34558	
585	248.894520	-0.452329	18.73576	16.83826	15.94847	15.53803	15.15638	
301	185.379456	0.128519	19.25762	19.24938	18.97530	18.88568	18.91675	
303	185.485219	0.163400	18.71479	16.97454	16.38521	16.12513	15.90649	
564	242.954823	-0.446457	18.90115	17.03906	16.03109	15.48150	15.06539	
75	160.938240	-0.423247	17.37104	15.78487	14.98982	14.58674	14.25872	
221	110.455531	1.254180	19.58684	17.73149	16.96642	16.67645	16.49140	
217	110.070039	1.218735	19.42591	18.42443	18.00135	17.79079	17.70945	
509	221.984190	0.446387	17.50681	16.05808	15.52935	15.36461	15.28286	
462	162.886970	-1.002881	17.79835	16.35954	15.67576	15.28575	14.98440	
453	162.766420	-0.860514	17.67043	16.20217	15.63600	15.44809	15.32646	
468	163.062060	-0.955493	18.42401	16.48616	15.58818	15.16028	14.82537	
85	161.094801	-0.452050	17.82710	16.23793	15.67866	15.49427	15.41664	
574	244.935545	-0.479247	18.53797	17.16947	16.50210	16.18933	15.95892	
434	154.611918	-0.164200	18.79575	18.57321	18.55136	18.35209	18.30321	
321	186.033836	0.113213	19.10102	17.89368	17.17626	16.78514	16.47559	
446	155.248147	-0.176081	18.43279	16.62750	15.71816	15.25928	14.88557	
567	244.112957	-0.492485	17.34144	17.66519	18.08743	18.37465	18.64286	
170	220.711731	0.461339	19.26163	17.71671	17.12826	16.89513	16.79474	
72	160.820289	-0.522241	16.34400	14.85487	14.29812	14.12055	14.08315	
277	207.456735	-0.726105	19.19459	18.31113	18.09644	17.83205	17.75919	
	run	camcol	field	specobjid	class	redshift	plate	mjd \
585	745.0	2	602	3.920000e+17	GALAXY	0.070187	348	51671
89	756.0	2	302	3.100000e+17	GALAXY	0.060674	275	51910
544	752.0	5	530	3.280000e+18	STAR	0.000026	2909	54653
464	756.0	1	315	3.110000e+17	QSO	0.358947	276	51909

105	756.0	1	304	2.690000e+18	STAR	0.000366	2389	54213
288	752.0	2	438	3.390000e+17	GALAXY	0.079088	301	51942
182	752.0	5	518	3.300000e+18	STAR	0.000032	2934	54626
472	756.0	1	316	3.110000e+17	STAR	0.000115	276	51909
156	752.0	5	508	3.460000e+17	QSO	1.257948	307	51663
218	308.0	6	26	8.390000e+18	STAR	0.000391	7450	56722
108	756.0	1	304	2.690000e+18	STAR	0.000239	2389	54213
447	756.0	1	313	2.880000e+18	STAR	0.001076	2559	54208
243	745.0	1	576	3.900000e+17	QSO	1.496161	346	51693
51	756.0	2	294	3.090000e+17	GALAXY	0.023351	274	51913
543	752.0	5	530	3.280000e+18	STAR	0.000123	2909	54653
578	745.0	2	583	4.100000e+17	GALAXY	0.093628	364	52000
239	745.0	1	568	3.900000e+17	QSO	1.062840	346	51693
163	752.0	5	511	3.460000e+17	GALAXY	0.037176	307	51663
297	752.0	2	443	3.390000e+17	GALAXY	0.106320	301	51942
178	752.0	5	518	3.460000e+17	STAR	0.000050	307	51663
162	752.0	5	510	3.460000e+17	GALAXY	0.116515	307	51663
532	752.0	5	527	3.480000e+17	GALAXY	0.040417	309	51994
25	752.0	4	274	3.240000e+17	GALAXY	0.116783	288	52000
414	756.0	3	251	3.040000e+17	GALAXY	0.088731	270	51909
375	756.0	2	308	2.880000e+18	STAR	-0.000076	2559	54208
29	752.0	4	275	2.880000e+18	STAR	0.000520	2558	54140
267	752.0	2	422	3.380000e+17	GALAXY	0.047497	300	51943
467	756.0	1	315	2.880000e+18	STAR	-0.000051	2559	54208
234	745.0	1	103	3.180000e+17	STAR	0.001135	282	51658
27	752.0	4	275	2.880000e+18	STAR	0.000175	2558	54140
..	...	...	...	...	...	...	...	...
553	745.0	1	591	4.100000e+17	STAR	0.000009	364	52000
21	752.0	4	274	3.240000e+17	STAR	-0.000036	288	52000
159	752.0	5	508	3.460000e+17	STAR	-0.000385	307	51663
134	756.0	1	310	2.690000e+18	STAR	-0.000456	2389	54213
158	752.0	5	508	4.530000e+18	STAR	-0.000276	4025	55350
499	752.0	5	522	4.530000e+18	GALAXY	0.055807	4022	55352
188	752.0	5	520	3.280000e+18	STAR	0.000836	2909	54653
222	308.0	6	28	8.390000e+18	STAR	0.000182	7450	56722
399	756.0	3	244	3.040000e+17	GALAXY	0.095463	270	51909
53	756.0	2	294	3.090000e+17	GALAXY	0.114168	274	51913
585	745.0	2	602	3.920000e+17	GALAXY	0.070187	348	51671
301	752.0	4	279	3.240000e+17	QSO	1.335835	288	52000
303	752.0	4	280	3.240000e+17	GALAXY	0.106216	288	52000
564	745.0	2	563	3.880000e+17	GALAXY	0.080116	345	51690
75	756.0	2	301	3.100000e+17	GALAXY	0.039109	275	51910
221	308.0	6	28	8.390000e+18	STAR	0.000346	7450	56722
217	308.0	6	25	8.390000e+18	STAR	0.000104	7450	56722
509	752.0	5	524	3.280000e+18	STAR	-0.000038	2909	54653
462	756.0	1	314	3.110000e+17	GALAXY	0.037305	276	51909
453	756.0	1	313	2.690000e+18	STAR	0.000153	2389	54213
468	756.0	1	315	3.110000e+17	GALAXY	0.048990	276	51909

85	756.0	2	302	2.880000e+18	STAR	-0.000078	2559	54208
574	745.0	2	576	3.900000e+17	GALAXY	0.059368	346	51693
434	756.0	3	259	3.050000e+17	QSO	1.812651	271	51883
321	752.0	4	284	3.250000e+17	GALAXY	0.156098	289	51990
446	756.0	3	263	3.050000e+17	GALAXY	0.055945	271	51883
567	745.0	2	570	3.900000e+17	STAR	-0.000005	346	51693
170	752.0	5	515	3.280000e+18	STAR	-0.000233	2909	54653
72	756.0	2	300	2.880000e+18	STAR	-0.000048	2559	54208
277	752.0	2	427	3.380000e+17	GALAXY	0.053106	300	51943

	fiberid
585	163
89	255
544	583
464	251
105	243
288	230
182	432
472	212
156	467
218	786
108	244
447	51
243	90
51	229
543	587
578	221
239	256
163	541
297	96
178	626
162	508
532	341
25	437
414	119
375	164
29	349
267	255
467	1
234	94
27	358
..	...
553	15
21	430
159	468
134	90
158	920
499	880

188	425
222	830
399	434
53	186
585	163
301	547
303	599
564	112
75	263
221	829
217	705
509	488
462	245
453	11
468	211
85	267
574	108
434	152
321	396
446	79
567	195
170	343
72	313
277	121

[100 rows x 16 columns]

```
In [103]: # two sided Hypothesis test
from scipy.stats import norm
from math import sqrt
```

```
def two_sided_hypo(sample_mean, pop_mean, std_dev, sample_size, alpha):
    actual_z = abs(norm.ppf(alpha/2))
    hypo_z = (sample_mean - pop_mean) / (std_dev/sqrt(sample_size))
    print('actual z value :', actual_z)
    print('hypothesis z value :', hypo_z, '\n')
    if hypo_z >= actual_z or hypo_z <= -(actual_z):
        return True
    else:
        return False
```

```
In [104]: #one sided hypothesis test(smaller than null hypothesis)
def one_sided_hypo(sample_mean, pop_mean, std_dev, sample_size, alpha):
    actual_z = abs(norm.ppf(alpha))
    hypo_z = (sample_mean - pop_mean) / (std_dev/sqrt(sample_size))
    print('actual z value :', actual_z)
    print('hypothesis z value :', hypo_z, '\n')
```

```

        if hypo_z >= actual_z:
            return True
        else:
            return False

```

In [105]: *# two sided hypothesis test*

```

alpha = 0.05
sample_mean = sample['ra'].mean()
pop_mean = df1['ra'].mean()
sample_size = 100
std_dev = df1['ra'].std()

print('H0 :  $\mu$  =', pop_mean)
print('H1 :  $\mu$  !=', pop_mean)
print('alpha value is :', alpha, '\n')

reject = two_sided_hypo(sample_mean, pop_mean, std_dev, sample_size, alpha)
if reject:
    print('Reject NULL hypothesis')
else:
    print('Failed to reject NULL hypothesis')

```

H0 :  $\mu$  = 185.01467401969944

H1 :  $\mu$  != 185.01467401969944

alpha value is : 0.05

actual z value : 1.9599639845400545

hypothesis z value : 2.1899914168152192

Reject NULL hypothesis

In [106]: *# one sided hypothesis test*

In [107]: alpha = 0.05

```

sample_mean = sample['u'].mean()
pop_mean = df1['u'].mean()
sample_size = 100
std_dev = df1['u'].std()

print('H0 :  $\mu$  <=', pop_mean)
print('H1 :  $\mu$  >', pop_mean)
print('alpha value is :', alpha, '\n')

reject = one_sided_hypo(sample_mean, pop_mean, std_dev, sample_size, alpha)
if reject:
    print('Reject NULL hypothesis')
else:
    print('Failed to reject NULL hypothesis')

```

H0 :  $\mu \leq 18.540124941569278$   
H1 :  $\mu > 18.540124941569278$   
alpha value is : 0.05

actual z value : 1.6448536269514729  
hypothesis z value : -0.9591187844229192

Failed to reject NULL hypothesis

In [108]: # 5) correlation

In [109]: # columns which are most and least correlated based on correlation co-efficient

In [113]: correlation\_matrix = df1.corr(method='pearson', min\_periods=2)

In [157]: #del correlation\_matrix['specobjid']  
correlation\_matrix.drop(['specobjid'], inplace = True)  
correlation\_matrix

Out[157]:

	ra	dec	u	g	r	i	\
ra	1.000000	0.010176	0.015354	0.078481	0.102108	0.118357	
dec	0.010176	1.000000	-0.004955	-0.028665	-0.017594	0.014852	
u	0.015354	-0.004955	1.000000	0.870558	0.726617	0.634071	
g	0.078481	-0.028665	0.870558	1.000000	0.962313	0.887921	
r	0.102108	-0.017594	0.726617	0.962313	1.000000	0.958512	
i	0.118357	0.014852	0.634071	0.887921	0.958512	1.000000	
z	0.114417	0.015491	0.595036	0.882497	0.968352	0.963430	
run	0.431633	-0.511464	0.018972	0.073280	0.074058	0.065505	
camcol	0.110580	0.978189	-0.008179	-0.012908	0.006059	0.040259	
field	0.879565	-0.093350	-0.005122	0.042027	0.063001	0.082867	
redshift	0.039125	-0.081308	0.136255	0.350250	0.386455	0.366967	
plate	-0.260961	0.376920	-0.156316	-0.151644	-0.090788	-0.039513	
mjd	-0.245881	0.310098	-0.207585	-0.181252	-0.104359	-0.047218	
fiberid	0.002694	0.712512	0.001515	0.010551	0.024853	0.049301	

	z	run	camcol	field	redshift	plate	\
ra	0.114417	0.431633	0.110580	0.879565	0.039125	-0.260961	
dec	0.015491	-0.511464	0.978189	-0.093350	-0.081308	0.376920	
u	0.595036	0.018972	-0.008179	-0.005122	0.136255	-0.156316	
g	0.882497	0.073280	-0.012908	0.042027	0.350250	-0.151644	
r	0.968352	0.074058	0.006059	0.063001	0.386455	-0.090788	
i	0.963430	0.065505	0.040259	0.082867	0.366967	-0.039513	
z	1.000000	0.058995	0.042313	0.073750	0.365302	-0.001333	
run	0.058995	1.000000	-0.390660	0.460247	0.068438	-0.454926	
camcol	0.042313	-0.390660	1.000000	-0.023134	-0.064613	0.344945	
field	0.073750	0.460247	-0.023134	1.000000	0.022774	-0.256309	
redshift	0.365302	0.068438	-0.064613	0.022774	1.000000	-0.150421	
plate	-0.001333	-0.454926	0.344945	-0.256309	-0.150421	1.000000	

```

mjd      -0.003059 -0.342423  0.292568 -0.223548 -0.159006  0.972875
fiberid   0.033802 -0.323926  0.721107 -0.110965 -0.034570  0.419519

```

```

           mjd  fiberid
ra      -0.245881  0.002694
dec       0.310098  0.712512
u       -0.207585  0.001515
g       -0.181252  0.010551
r       -0.104359  0.024853
i       -0.047218  0.049301
z       -0.003059  0.033802
run      -0.342423 -0.323926
camcol    0.292568  0.721107
field    -0.223548 -0.110965
redshift -0.159006 -0.034570
plate     0.972875  0.419519
mjd       1.000000  0.360050
fiberid   0.360050  1.000000

```

```

In [144]: # Finding the most correlated values of each column based on pearson correlation coeff
          correlation_matrix['ra'].nlargest(2)

```

```

Out[144]: ra      1.000000
          field    0.879565
          Name: ra, dtype: float64

```

```

In [117]: correlation_matrix['dec'].nlargest(2)

```

```

Out[117]: dec      1.000000
          camcol    0.978189
          Name: dec, dtype: float64

```

```

In [118]: correlation_matrix['u'].nlargest(2)

```

```

Out[118]: u      1.000000
          g      0.870558
          Name: u, dtype: float64

```

```

In [119]: correlation_matrix['g'].nlargest(2)

```

```

Out[119]: g      1.000000
          r      0.962313
          Name: g, dtype: float64

```

```

In [120]: correlation_matrix['r'].nlargest(2)

```

```

Out[120]: r      1.000000
          z      0.968352
          Name: r, dtype: float64

```



```

In [121]: correlation_matrix['i'].nlargest(2)

Out[121]: i      1.00000
         z      0.96343
         Name: i, dtype: float64

In [122]: correlation_matrix['z'].nlargest(2)

Out[122]: z      1.000000
         r      0.968352
         Name: z, dtype: float64

In [123]: correlation_matrix['run'].nlargest(2)

Out[123]: run      1.000000
         field     0.460247
         Name: run, dtype: float64

In [124]: correlation_matrix['camcol'].nlargest(2)

Out[124]: camcol     1.000000
         dec         0.978189
         Name: camcol, dtype: float64

In [125]: correlation_matrix['field'].nlargest(2)

Out[125]: field     1.000000
         ra         0.879565
         Name: field, dtype: float64

In [131]: correlation_matrix['redshift'].nlargest(2)

Out[131]: redshift     1.000000
         r             0.386455
         Name: redshift, dtype: float64

In [132]: # We can see that camcol and dec are most correlated
          # The type of camera scanline used highly effects the declination angle to a large ext

          # Second most correlated are g,r,i,z as all of this are color bands responsiveness of
          # where g,r,i,z are the wavelenths g - green , r - red , i - infrared , z - 900nm
          # hence the responsiveness of the telescope is same for all the wavelenghts measured

In [158]: # Finding the least correlated values of each column based on pearson correlation coff
          correlation_matrix['ra'].nsmallest(2)

Out[158]: plate     -0.260961
         mjd        -0.245881
         Name: ra, dtype: float64

```

```

In [159]: correlation_matrix['dec'].nsmallest(2)

Out[159]: run      -0.511464
          field    -0.093350
          Name: dec, dtype: float64

In [160]: correlation_matrix['u'].nsmallest(2)

Out[160]: mjd      -0.207585
          plate    -0.156316
          Name: u, dtype: float64

In [161]: correlation_matrix['g'].nsmallest(2)

Out[161]: mjd      -0.181252
          plate    -0.151644
          Name: g, dtype: float64

In [162]: correlation_matrix['r'].nsmallest(2)

Out[162]: mjd      -0.104359
          plate    -0.090788
          Name: r, dtype: float64

In [135]: correlation_matrix['i'].nsmallest(2)

Out[135]: mjd      -0.047218
          specobjid -0.039527
          Name: i, dtype: float64

In [163]: correlation_matrix['z'].nsmallest(2)

Out[163]: mjd      -0.003059
          plate    -0.001333
          Name: z, dtype: float64

In [164]: correlation_matrix['run'].nsmallest(2)

Out[164]: dec      -0.511464
          plate    -0.454926
          Name: run, dtype: float64

In [165]: correlation_matrix['camcol'].nsmallest(2)

Out[165]: run      -0.390660
          redshift  -0.064613
          Name: camcol, dtype: float64

In [166]: correlation_matrix['field'].nsmallest(2)

```

```
Out[166]: plate    -0.256309
          mjd      -0.223548
          Name: field, dtype: float64
```

```
In [167]: correlation_matrix['redshift'].nsmallest(2)
```

```
Out[167]: mjd      -0.159006
          plate    -0.150421
          Name: redshift, dtype: float64
```

```
In [168]: # We see that the least correlated features are the dec and the run
          # As the declination angle does not depend on the type of the scanline we use in the t

          # We see that the second least correlated features are the dec and the plate
          # As the declination angle does not depend on the plate used in the telescope
```