

# kepler-dataset-exploratory-analysis

February 6, 2022

## 1 Kepler candidates dataset Exploratory Data Analysis.

```
[1]: import pandas as pd
import matplotlib.pyplot as plt

%matplotlib inline
```

```
[2]: data = pd.read_csv('cumulative.csv')
```

```
[3]: data.head()
```

```
[3]:
```

	rowid	kepid	kepoi_name	kepler_name	koi_disposition	koi_pdisposition	\
0	1	10797460	K00752.01	Kepler-227 b	CONFIRMED	CANDIDATE	
1	2	10797460	K00752.02	Kepler-227 c	CONFIRMED	CANDIDATE	
2	3	10811496	K00753.01	NaN	FALSE POSITIVE	FALSE POSITIVE	
3	4	10848459	K00754.01	NaN	FALSE POSITIVE	FALSE POSITIVE	
4	5	10854555	K00755.01	Kepler-664 b	CONFIRMED	CANDIDATE	

	koi_score	koi_fpflag_nt	koi_fpflag_ss	koi_fpflag_co	...	\
0	1.000	0	0	0	...	
1	0.969	0	0	0	...	
2	0.000	0	1	0	...	
3	0.000	0	1	0	...	
4	1.000	0	0	0	...	

	koi_steff_err2	koi_slogg	koi_slogg_err1	koi_slogg_err2	koi_srad	\
0	-81.0	4.467	0.064	-0.096	0.927	
1	-81.0	4.467	0.064	-0.096	0.927	
2	-176.0	4.544	0.044	-0.176	0.868	
3	-174.0	4.564	0.053	-0.168	0.791	
4	-211.0	4.438	0.070	-0.210	1.046	

	koi_srad_err1	koi_srad_err2	ra	dec	koi_kepmag
0	0.105	-0.061	291.93423	48.141651	15.347
1	0.105	-0.061	291.93423	48.141651	15.347
2	0.233	-0.078	297.00482	48.134129	15.436
3	0.201	-0.067	285.53461	48.285210	15.597

```
4          0.334          -0.133  288.75488  48.226200          15.509
```

```
[5 rows x 50 columns]
```

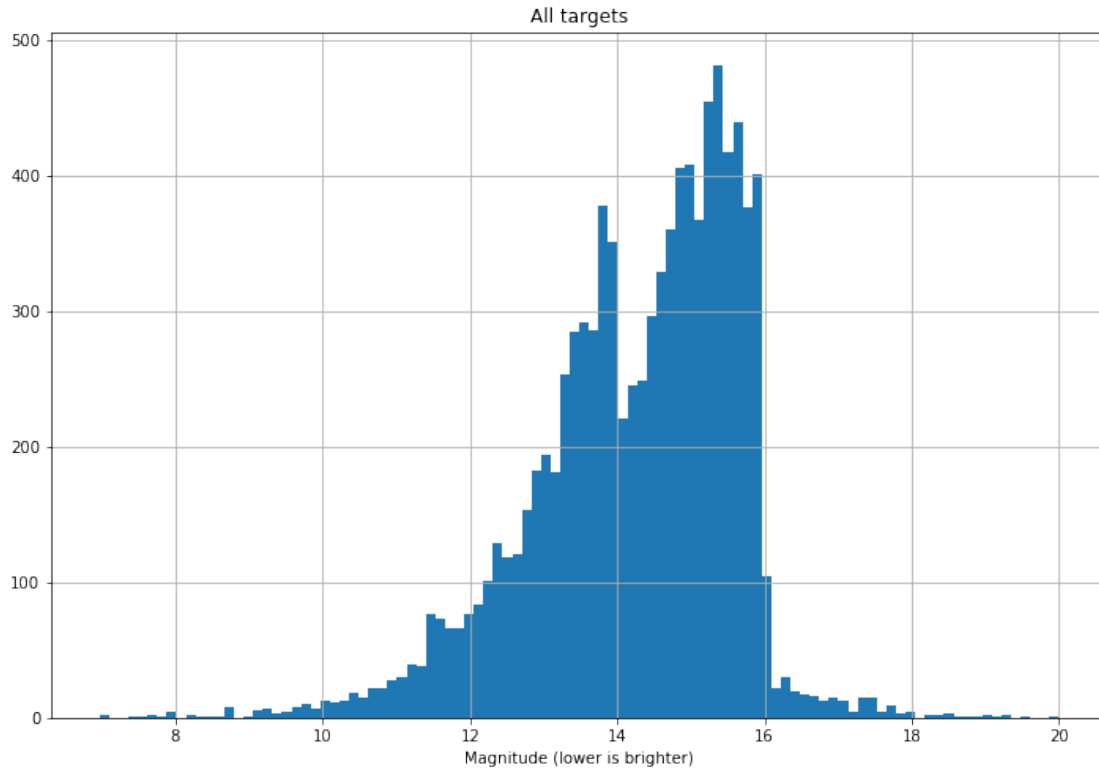
```
[4]: data.columns
```

```
[4]: Index(['rowid', 'kepid', 'kepoi_name', 'kepler_name', 'koi_disposition',
        'koi_pdisposition', 'koi_score', 'koi_fpflag_nt', 'koi_fpflag_ss',
        'koi_fpflag_co', 'koi_fpflag_ec', 'koi_period', 'koi_period_err1',
        'koi_period_err2', 'koi_time0bk', 'koi_time0bk_err1',
        'koi_time0bk_err2', 'koi_impact', 'koi_impact_err1', 'koi_impact_err2',
        'koi_duration', 'koi_duration_err1', 'koi_duration_err2', 'koi_depth',
        'koi_depth_err1', 'koi_depth_err2', 'koi_prad', 'koi_prad_err1',
        'koi_prad_err2', 'koi_teq', 'koi_teq_err1', 'koi_teq_err2', 'koi_insol',
        'koi_insol_err1', 'koi_insol_err2', 'koi_model_snr', 'koi_tce_plnt_num',
        'koi_tce_delivname', 'koi_steff', 'koi_steff_err1', 'koi_steff_err2',
        'koi_slogg', 'koi_slogg_err1', 'koi_slogg_err2', 'koi_srad',
        'koi_srad_err1', 'koi_srad_err2', 'ra', 'dec', 'koi_kepmag'],
        dtype='object')
```

## 1.1 Figures

```
[5]: ax = data.koi_kepmag.hist(bins=100, figsize=(12, 8))
      ax.set_xlabel("Magnitude (lower is brighter)")
      ax.set_title("All targets")
```

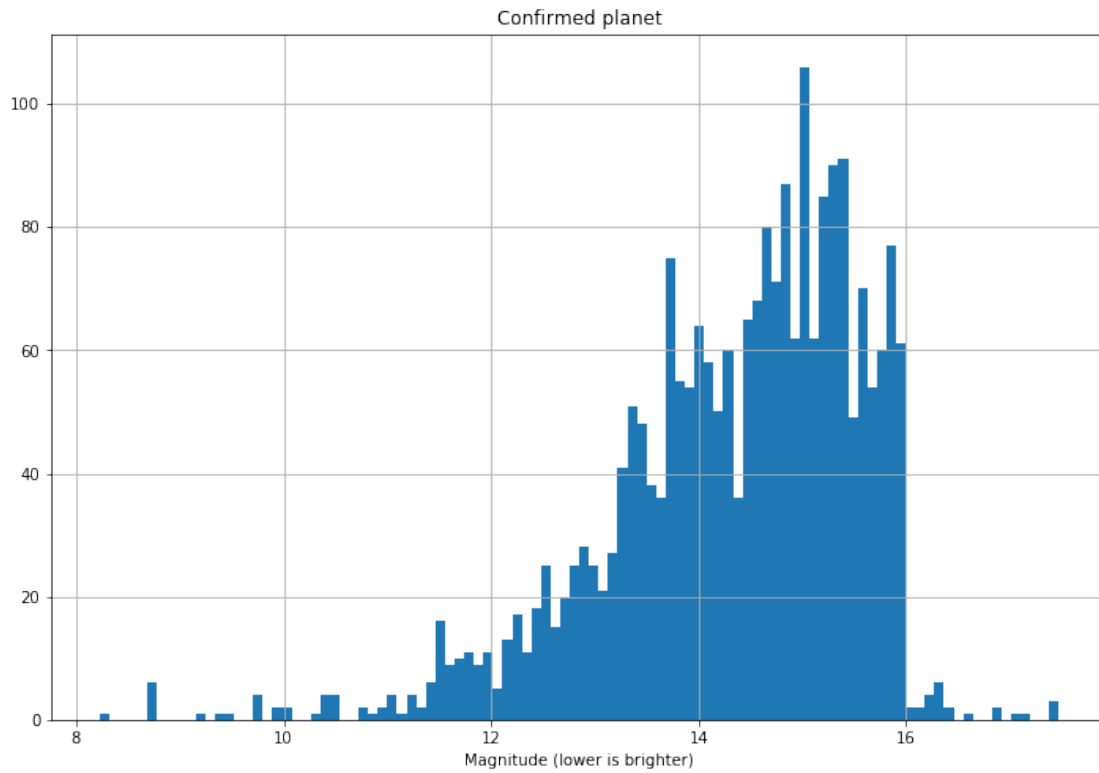
```
[5]: <matplotlib.text.Text at 0x7f0ef05d8fd0>
```



We see what looks like a bi-modal distribution and a sharp cut-off at around 16th magnitude. This is likely the limit at which the noise becomes too much to detect planetary candidates.

```
[6]: # only the Confirmed planets
ax = data[data.koi_disposition == 'CONFIRMED'].koi_kepmag.hist(bins=100,
    figsize=(12, 8))
ax.set_xlabel("Magnitude (lower is brighter)")
ax.set_title("Confirmed planet")
```

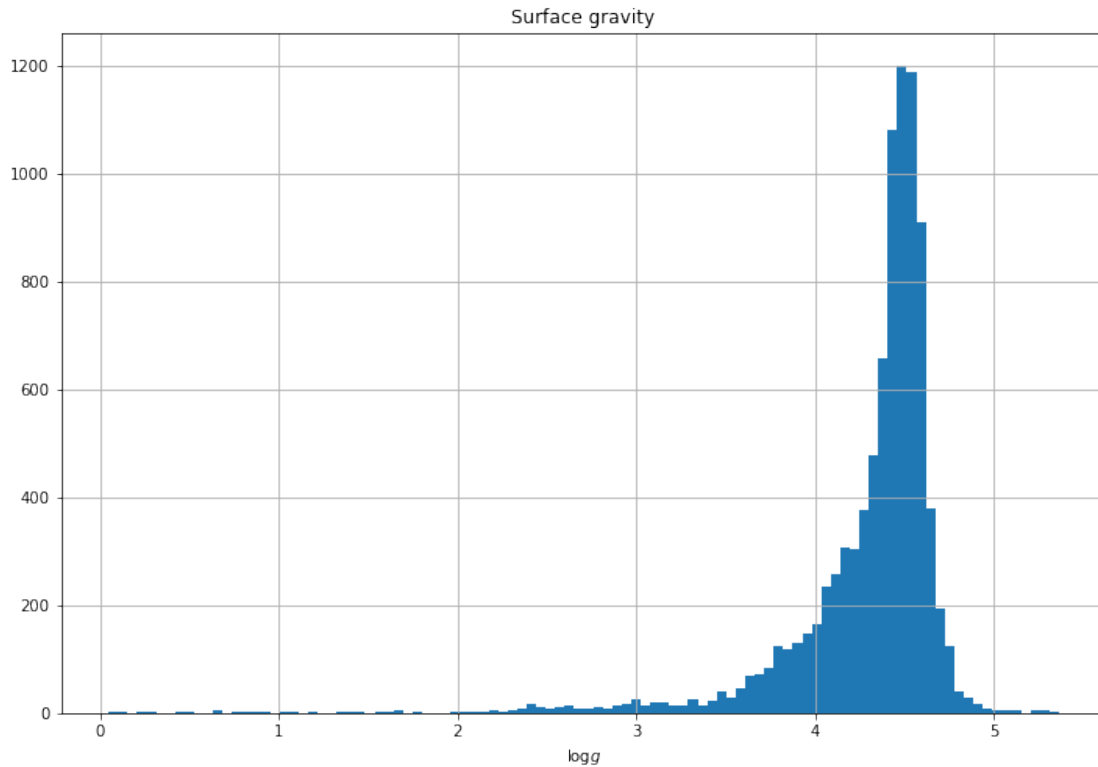
```
[6]: <matplotlib.text.Text at 0x7f0ef01d0a90>
```



### 1.1.1 Surface Gravity of Stars

```
[7]: ax = data.koi_slogg.hist(bins=100, figsize=(12, 8))  
ax.set_xlabel("$\\log{g}$")  
ax.set_title("Surface gravity")
```

```
[7]: <matplotlib.text.Text at 0x7f0eefdefda0>
```



### 1.1.2 Right Ascension and Declination

```
[8]: confirmed = data[data.koi_disposition == 'CONFIRMED']

ra, dec = data.ra, data.dec
ra_c, dec_c = confirmed.ra, confirmed.dec
```

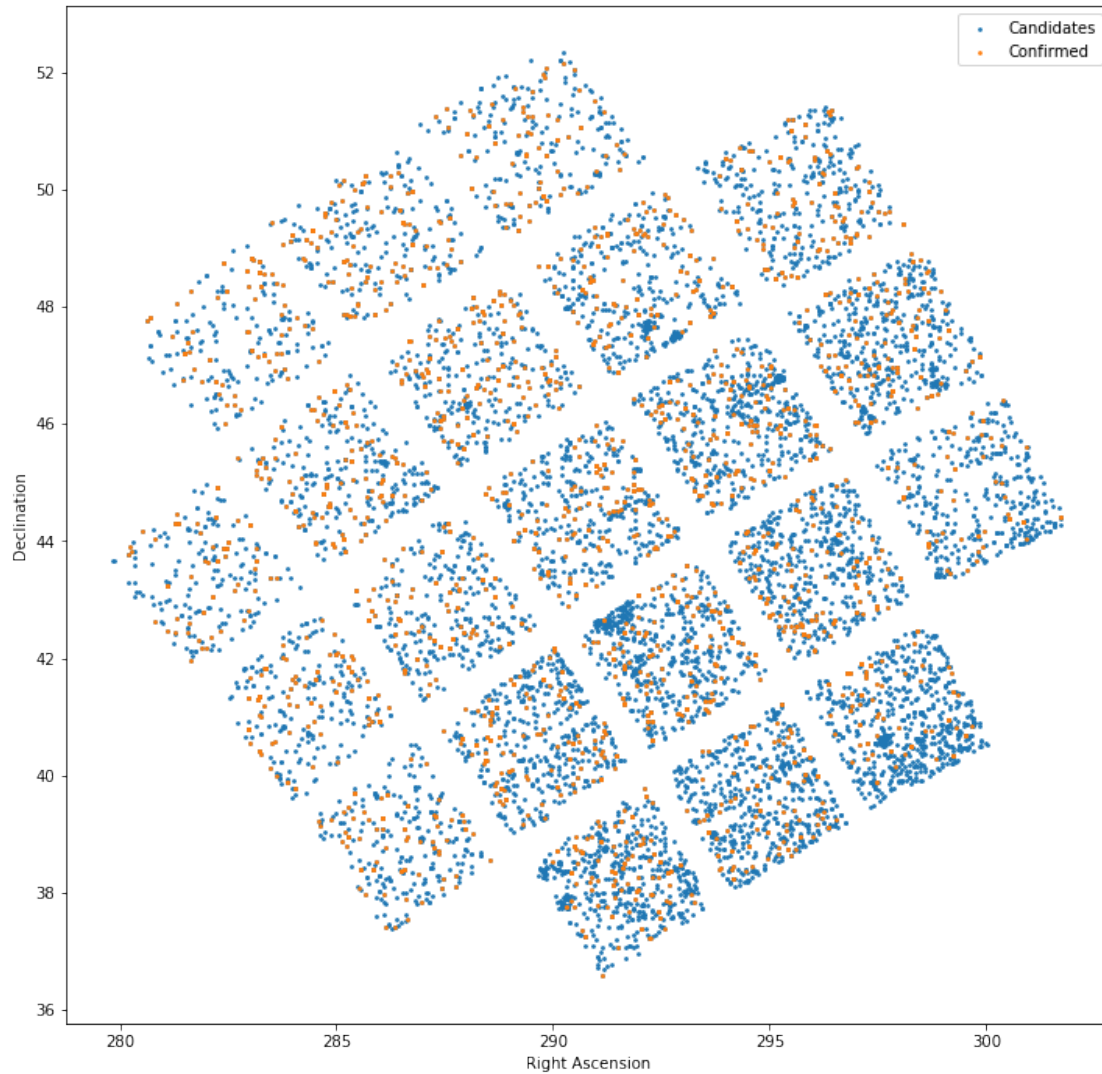
```
[9]: fig = plt.figure(figsize=(12, 12))

plt.scatter(ra, dec, s=3, label='Candidates')
plt.scatter(ra_c, dec_c, s=3, label="Confirmed")

plt.xlabel("Right Ascension")
plt.ylabel("Declination")

plt.legend()
```

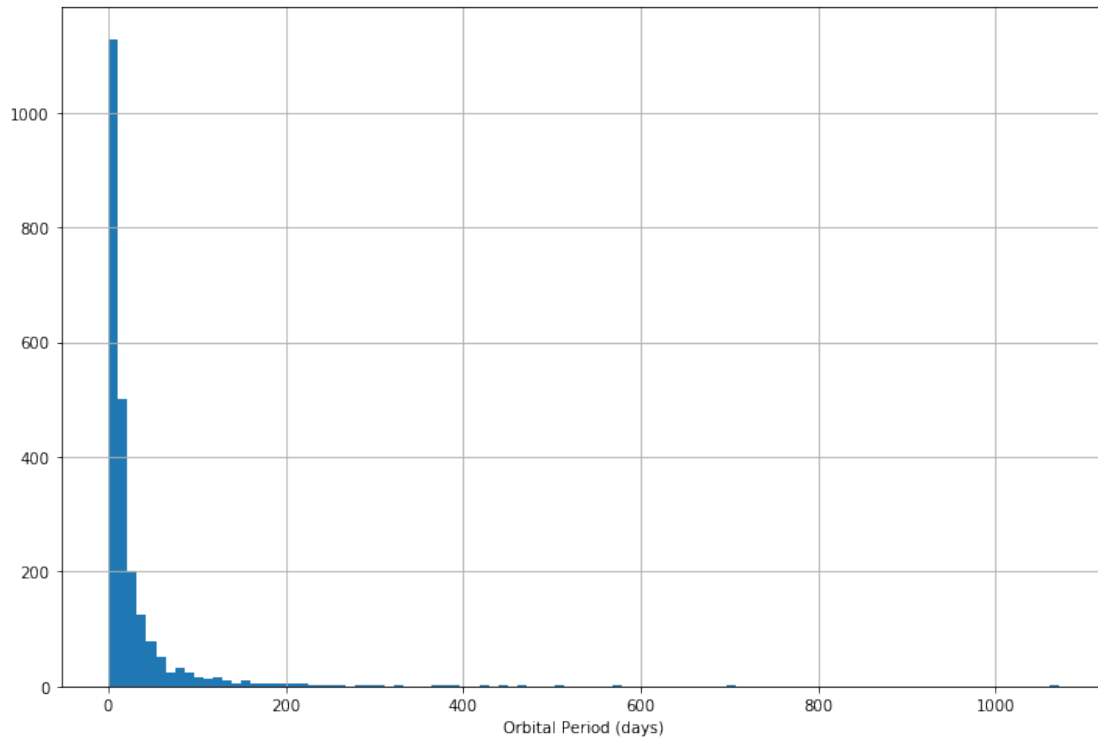
```
[9]: <matplotlib.legend.Legend at 0x7f0eef77cbe0>
```



### 1.1.3 Orbital periods of confirmed planets

```
[10]: ax = confirmed.koi_period.hist(bins=100, figsize=(12, 8))  
      ax.set_xlabel("Orbital Period (days)")
```

```
[10]: <matplotlib.text.Text at 0x7f0eef7aa358>
```

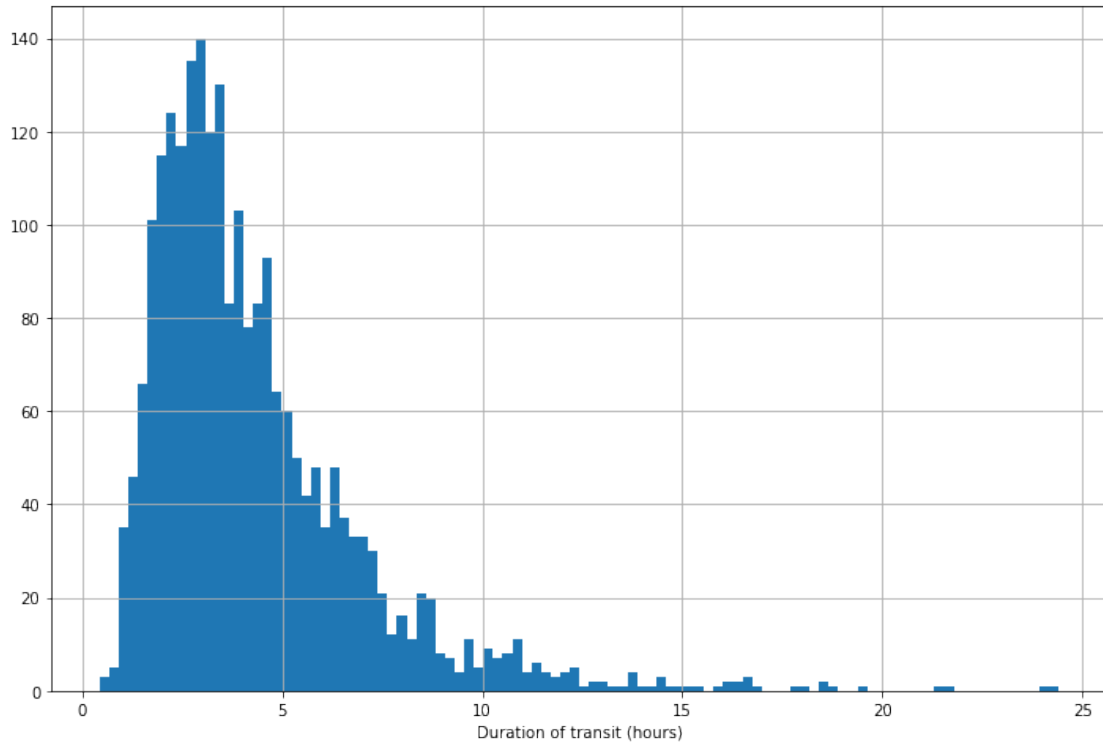


Most confirmed planets have lower orbital periods. This makes sense, the closer the planet is to its parent star, the more likely it is to eclipse it (and hence the more likely it is to be observed.)

#### 1.1.4 Duration of planetary transits

```
[11]: ax = confirmed.koi_duration.hist(bins=100, figsize=(12, 8))  
      ax.set_xlabel("Duration of transit (hours)")
```

```
[11]: <matplotlib.text.Text at 0x7f0eef694d30>
```



Durations are on the order of hours, with more detections at smaller hours, again showing the detection bias towards smaller orbits.

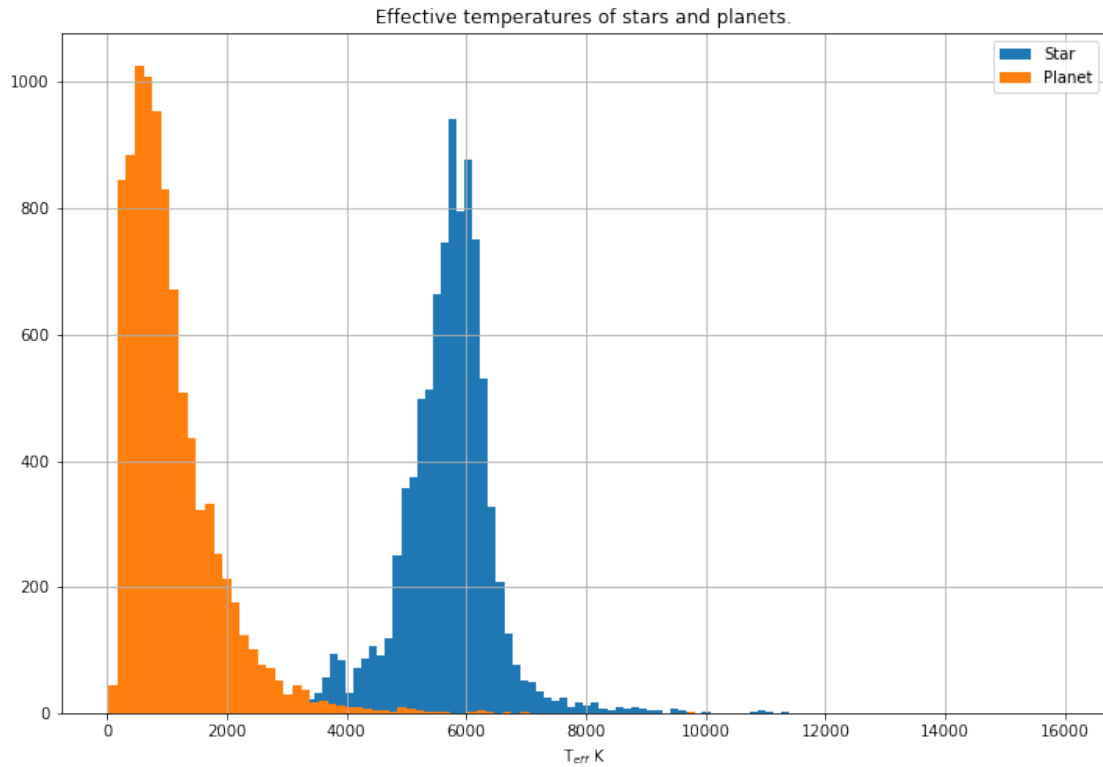
### 1.1.5 Temperatures and Radii of stars and planets

```
[12]: ax = data.koi_steff.hist(bins=100, figsize=(12, 8), label="Star")
ax.set_xlabel("T$_{eff}$ K")
ax.set_title("Effective temperatures of stars and planets.")

data.koi_teq.hist(ax=ax, bins=100, label='Planet')
ax.legend()
```

```
[12]: <matplotlib.legend.Legend at 0x7f0eef9af240>
```





```
[13]: # radius only for stars with a confirmed planet, the rest is not known.
ax = confirmed.koi_prad.hist(bins=100, label='Planet (earth radii)')
ax.set_title("Radii of stars and planets")

confirmed.koi_srad.hist(ax=ax, bins=100, figsize=(12, 8), label="Star (solar_
→radii)")
ax.set_xlabel("Radius")
ax.legend()
```

```
[13]: <matplotlib.legend.Legend at 0x7f0eed38ae48>
```

