

I first convert all fits files to csv. I took this step because csv is one of the most commonly known formats while fits is a format quite specific to the Astronomy community. Hence, for a general data scientist with no knowledge of the fits format, it will be easier to deal with a csv format as it can be seen and modified with tools available in almost every operating system (while fits requires a specific fits viewer). In addition, sqlite cannot deal with data provided with fits format while it is extremely easy to deal with csv in sqlite by just putting .mode csv.

To make this pre-processing formal, I have provided the python file named fitstocsv.py. It can be modified and used by any other person in a similar situation. I also have another python file that adds two columns named 'MJD' and 'FieldID' with their values to these csv files to make another csv file of the format *-all.csv. the purpose of this will be clear when I later use these columns as foreign keys.

NOTE ON 23rd January: I missed the fact that the Ks observations are listed in order. So, I made an ordering of the different observation files based on the date and time in the filename in the File info csv file. This is exactly what you had mentioned not to do and I somehow missed that. I realized the correct order much later (today in fact, when I've done all the questions for about a week now) but correcting it now is going to take a lot of time and effort (it will affect almost all operations that follow) but right now I'm constrained by my First Year Master's Project. So, I apologize for this transgression and me being unable to correct it. My convention is as follows:

FieldID	Filename	Filter	MJD
1	Field-1-Z-all.csv	Z	57267.1671072
1	Field-1-J-all.csv	J	57257.0504323
1	Field-1-H-all.csv	H	57257.044108
1	Field-1-Ks-E001-all.csv	Ks	56788.346937
1	Field-1-Ks-E003-all.csv	Ks	56561.0020158
1	Field-1-Ks-E002-all.csv	Ks	56829.0390512
1	Field-1-Y-all.csv	Y	57267.1596647
2	Field-2-Z-all.csv	Z	57268.1671072
2	Field-2-J-all.csv	J	57258.0504323
2	Field-2-H-all.csv	H	57258.044108
2	Field-2-Ks-E001-all.csv	Ks	56789.346937
2	Field-2-Y-all.csv	Y	57268.1596647
3	Field-3-Z-all.csv	Z	57268.1671072
3	Field-3-J-all.csv	J	57258.0504323
3	Field-3-H-all.csv	H	57258.044108
3	Field-3-Ks-E001-all.csv	Ks	56789.346937
3	Field-3-Ks-E002-all.csv	Ks	56562.0020158
3	Field-3-Y-all.csv	Y	57268.1596647

These are also saved in the tabledata\locationq1.csv which forms the data for Relationtable in test2.db.

1a)

The details of the question leads to the conclusion that the database should be a 'Relational Database' because the table with the locations of the data files and the data files themselves need to be linked by foreign keys. Hence, it will be good to import all data provided to us as separate tables as well as make another table which in turn links to these data tables.

The data tables are named Stars(Field)(Filter)(Additional Number if Ks filter here named K). The table which in turn is linked to all these tables by foreign keys is named as Relationtable. The foreign keys link the Field ID and MJD of this table to the two columns named FieldID and MJD in each data table. It is important to note that here RElationtable is the child to all data tables. So, a new data can be added to the database only when the constraint laid by the foreign keys are followed. A schema for creation of this database is provided in createtablefromcsv.sql.

Now the database test2.db is imported by running `sqlite3 test2.db` on the terminal. Then the .sql script is run as `.run createtablefromcsv.sql`. I have already done this, so the tables can simply be seen by typing `.tables`. Now, we are ready to formulate sql queries to get relevant results.

1b)

Note: Signal is taken to be Flux1 and Noise as dFlux1, Colour (Y,J,H,Ks,Z) is taken to be Mag1.

R1a:

All images observed between MJD=56800 and MJD=57300

Sql query:

```
SELECT Filename FROM Relationtable
WHERE MJD<57300.0 AND MJD>56800.0;
```

Result of query:

Field-1-Z-all.csv
Field-1-J-all.csv
Field-1-H-all.csv
Field-1-Ks-E002-all.csv
Field-1-Y-all.csv
Field-2-Z-all.csv
Field-2-J-all.csv
Field-2-H-all.csv
Field-2-Y-all.csv
Field-3-Z-all.csv
Field-3-J-all.csv
Field-3-H-all.csv
Field-3-Y-all.csv

13 out of 18 data files fulfil the criterion

R1b:

The number of stars detected with $S/N > 5$ in each image

Sql query:

```
SELECT COUNT(*) as 'NStars' FROM Stars1Z
WHERE Flux1/dFlux1>5;
SELECT COUNT(*) as 'NStars' FROM Stars1J
WHERE Flux1/dFlux1>5;
SELECT COUNT(*) as 'NStars' FROM Stars1H
WHERE Flux1/dFlux1>5;
SELECT COUNT(*) as 'NStars' FROM Stars1K2
```

```

WHERE Flux1/dFlux1>5;
SELECT COUNT(*) as 'NStars' FROM Stars1Y
WHERE Flux1/dFlux1>5;
SELECT COUNT(*) as 'NStars' FROM Stars2Z
WHERE Flux1/dFlux1>5;
SELECT COUNT(*) as 'NStars' FROM Stars2J
WHERE Flux1/dFlux1>5;
SELECT COUNT(*) as 'NStars' FROM Stars2H
WHERE Flux1/dFlux1>5;
SELECT COUNT(*) as 'NStars' FROM Stars2Y
WHERE Flux1/dFlux1>5;
SELECT COUNT(*) as 'NStars' FROM Stars3Z
WHERE Flux1/dFlux1>5;
SELECT COUNT(*) as 'NStars' FROM Stars3J
WHERE Flux1/dFlux1>5;
SELECT COUNT(*) as 'NStars' FROM Stars3H
WHERE Flux1/dFlux1>5;
SELECT COUNT(*) as 'NStars' FROM Stars3Y
WHERE Flux1/dFlux1>5;

```

Result on query (in respective order):

```

9952
10000
10000
9997
9998
9992
10000
9996
10000
9990
10000
10000
10000

```

R2:

Find the objects that have $J-H > 1.5$

SQL query:

```

.mode csv
.output JHfield1.csv
SELECT s.StarID as n, s.Mag1-o.Mag1 as JH
FROM
    Stars1J as s
    JOIN Stars1H as o
    ON n=o.StarID
WHERE JH > 1.5;

```

```

.mode csv
.output JHfield2.csv

```

```

SELECT s.StarID as n, s.Mag1-o.Mag1 as JH
FROM
    Stars2J as s
    JOIN Stars2H as o
    ON n=o.StarID
WHERE JH > 1.5;

```

```

.mode csv
.output JHfield3.csv
SELECT s.StarID as n, s.Mag1-o.Mag1 as JH
FROM
    Stars3J as s
    JOIN Stars3H as o
    ON n=o.StarID
WHERE JH > 1.5;

```

Query Result saved as csv files(respectively) and histogram plots are drawn using the python file histogram.py:

JHfield1.csv
 JHfield2.csv
 JHfield3.csv

Comment on the graphs made:

As expected the graphs start from J-H = 1.5. The number of stars seem to follow an exponentially decreasing distribution after that. Also, Field 1 seems to have a overall larger number of stars overall while Field 2 seems to have the overall least for this criterion.

R3:

Find the objects where Ks differs by more than 20 times the flux uncertainty from the mean flux.

Assumption:

This is a bit vague. The way I understand it, this may mean finding the difference between Mag1 and Flux1 and checking if this exceeds twenty times the value of dFlux1. Since, the LHS isn't an absolute value, it could either be Mag1-Flux1 or Flux1-Mag1. The query for the former is provided below.

SQL query:

```

SELECT STARID as n, Mag1 as K FROM Stars1K1
WHERE (Mag1-Flux1)>(20.0*dFlux1);

```

```

SELECT STARID as n, Mag1 as K FROM Stars1K2
WHERE (Mag1-Flux1)>(20.0*dFlux1);

```

```

SELECT STARID as n, Mag1 as K FROM Stars1K3
WHERE (Mag1-Flux1)>(20.0*dFlux1);

```

```

SELECT STARID as n, Mag1 as K FROM Stars2K1
WHERE (Mag1-Flux1)>(20.0*dFlux1);

```

```

SELECT STARID as n, Mag1 as K FROM Stars3K1
WHERE (Mag1-Flux1)>(20.0*dFlux1);

```

```
SELECT STARID as n, Mag1 as K FROM Stars3K2
WHERE (Mag1-Flux1)>(20.0*dFlux1);
```

Result of Respective queries:

None

None

None

None

None

301288|15.0318

For the later case,

SQL query:

.mode csv

.output FMdF1K1.csv

```
SELECT STARID as n, Mag1 as K FROM Stars1K1
WHERE (Flux1-Mag1)>(20.0*dFlux1);
```

.mode csv

.output FMdF1K2.csv

```
SELECT STARID as n, Mag1 as K FROM Stars1K2
WHERE (Flux1-Mag1)>(20.0*dFlux1);
```

.mode csv

.output FMdF1K3.csv

```
SELECT STARID as n, Mag1 as K FROM Stars1K3
WHERE (Flux1-Mag1)>(20.0*dFlux1);
```

.mode csv

.output FMdF2K1.csv

```
SELECT STARID as n, Mag1 as K FROM Stars2K1
WHERE (Flux1-Mag1)>(20.0*dFlux1);
```

.mode csv

.output FMdF3K1.csv

```
SELECT STARID as n, Mag1 as K FROM Stars3K1
WHERE (Flux1-Mag1)>(20.0*dFlux1);
```

.mode csv

.output FMdF3K2.csv

```
SELECT STARID as n, Mag1 as K FROM Stars3K2
WHERE (Flux1-Mag1)>(20.0*dFlux1);
```

Results of queries are stored in the respective csv files mentioned as .output. The data from these files is made into histograms with the help of TOPCAT (bandwidth = 0.1) with format (FieldID)K(Observation number).png.

Comments on the images:

It is clear that different observations in the same field give different numbers of stars overall. 1K1

has the least amount of stars and 1K3 has the most in Field 1. Even the shape of the distribution in 1K3 is different (bimodal) than the rest two in Field 1. The difference in distribution is reduced in Field 3, however FK2 has more stars than FK1. In general, the distributions of all stars in different fields tends to be the same.

R4:

Find all catalogues that exist for a given field.

SQL query:

```
SELECT Filename FROM Relationtable
WHERE FieldID = 1;
```

```
SELECT Filename FROM Relationtable
WHERE FieldID = 2;
```

```
SELECT Filename FROM Relationtable
WHERE FieldID = 3;
```

Result of queries:

Field-1-Z-all.csv

Field-1-J-all.csv

Field-1-H-all.csv

Field-1-Ks-E001-all.csv

Field-1-Ks-E003-all.csv

Field-1-Ks-E002-all.csv

Field-1-Y-all.csv

Field-2-Z-all.csv

Field-2-J-all.csv

Field-2-H-all.csv

Field-2-Ks-E001-all.csv

Field-2-Y-all.csv

Field-3-Z-all.csv

Field-3-J-all.csv

Field-3-H-all.csv

Field-3-Ks-E001-all.csv

Field-3-Ks-E002-all.csv

Field-3-Y-all.csv

R5:

For a given field I would like to retrieve the Y, Z, J, H and Ks magnitudes for all stars with S/N > 30 in Y, Z, J, H and Ks.

SQL query:

.mode csv

.output allcoloursfield1.csv

```
SELECT a.starID as n, a.Mag1 as Y, b.Mag1 as Z, c.Mag1 as H, d.Mag1 as J, e.Mag1 as K1,
f.Mag1 as K2, g.Mag1 as K3
```

```
FROM
```

```
Stars1Y as a
```

```

JOIN STars1Z as b ON n=b.StarID
JOIN Stars1H as c ON n=c.StarID
JOIN Stars1J as d ON n=d.StarID
JOIN Stars1K1 as e ON n=e.StarID
JOIN Stars1K2 as f ON n=f.StarID
JOIN Stars1K3 as g ON n=g.StarID
WHERE a.Flux1/a.dFlux1 > 30 AND b.Flux1/b.dFlux1 > 30 AND c.Flux1/c.dFlux1 > 30 AND
d.Flux1/d.dFlux1 > 30 AND e.Flux1/e.dFlux1 > 30 AND f.Flux1/f.dFlux1 >30 AND
g.Flux1/g.dFlux1 > 30;

```

```

.mode csv
.output allcoloursfield2.csv
SELECT a.starID as n, a.Mag1 as Y, b.Mag1 as Z, c.Mag1 as H, d.Mag1 as J, e.Mag1 as K1
FROM
    Stars2Y as a
    JOIN STars2Z as b ON n=b.StarID
    JOIN Stars2H as c ON n=c.StarID
    JOIN Stars2J as d ON n=d.StarID
    JOIN Stars2K1 as e ON n=e.StarID
WHERE a.Flux1/a.dFlux1 > 30 AND b.Flux1/b.dFlux1 > 30 AND c.Flux1/c.dFlux1 > 30 AND
d.Flux1/d.dFlux1 > 30 AND e.Flux1/e.dFlux1 > 30;

```

```

.mode csv
.output allcoloursfield3.csv
SELECT a.starID as n, a.Mag1 as Y, b.Mag1 as Z, c.Mag1 as H, d.Mag1 as J, e.Mag1 as K1,
f.Mag1 as K2
FROM
    Stars3Y as a
    JOIN STars3Z as b ON n=b.StarID
    JOIN Stars3H as c ON n=c.StarID
    JOIN Stars3J as d ON n=d.StarID
    JOIN Stars3K1 as e ON n=e.StarID
    JOIN Stars3K2 as f ON n=f.StarID
WHERE a.Flux1/a.dFlux1 > 30 AND b.Flux1/b.dFlux1 > 30 AND c.Flux1/c.dFlux1 > 30 AND
d.Flux1/d.dFlux1 > 30 AND e.Flux1/e.dFlux1 > 30 AND f.Flux1/f.dFlux1 >30;

```

Results of queries are stored in the .csv files named in the queries. Histograms for each colour for each field is extracted using TOPCAT (bandwidth = 0.1) and saved as Field(FieldID)(Colour) (Observation number for K Colour).png.

1c)

Constructing a table of Stars with Y-J and J-H magnitudes. Thwn using this table to create a simulated list of 100,000 stars for the Euclid mission.

Assumption:

To do this exercise, I assume that the distribution of stars I get from the database belong to a sample that can be reconstructed from a Kernel Density Estimation to construct a probability density distribution. That would mean assuming these stars are a representative sample of the distribution. Also, I assume a minimum S/N ratio of 5 to assume good quality observations.

SQL query:

.mode csv

.headers on

.output YJHcatalog.csv

```
SELECT s.StarID as n, s.Mag1-o.Mag1 as YJ, o.Mag1-p.Mag1 as JH
FROM Stars1Y as s
      JOIN Stars1J as o ON n=o.StarID
      JOIN Stars1H as p ON o.StarID = p.StarID
WHERE s.Flux1/s.dFlux1 > 5 AND o.Flux1/o.dFlux1 > 5 AND p.Flux1/p.dFlux1 > 5
UNION
SELECT s.StarID as n, s.Mag1-o.Mag1 as YJ, o.Mag1-p.Mag1 as JH
FROM Stars2Y as s
      JOIN Stars2J as o ON n=o.StarID
      JOIN Stars2H as p ON o.StarID = p.StarID
WHERE s.Flux1/s.dFlux1 > 5 AND o.Flux1/o.dFlux1 > 5 AND p.Flux1/p.dFlux1 > 5
UNION
SELECT s.StarID as n, s.Mag1-o.Mag1 as YJ, o.Mag1-p.Mag1 as JH
FROM Stars3Y as s
      JOIN Stars3J as o ON n=o.StarID
      JOIN Stars3H as p ON o.StarID = p.StarID
WHERE s.Flux1/s.dFlux1 > 5 AND o.Flux1/o.dFlux1 > 5 AND p.Flux1/p.dFlux1 > 5;
```

Now the output catalog is used as an input to the Gaussian Kernel Density estimation done in q1_c_simulation.ipynb to create a simulated list of 100,000 stars. I use Gaussian Kernel because at the moment (sklearn restriction) random samples can be drawn from the distribution constructed using Gaussian and Tophat kernels. Gaussian kernel is in general more preferable for a large distribution like the one we're using.

References for codes used here and the .sql file:

<http://www.sqlite.org/foreignkeys.html>

<https://www.sqlservercentral.com/Forums/Topic1540629-3077-1.aspx>

<http://www.sqlitetutorial.net/sqlite-export-csv/>

<http://www.dofactory.com/sql/join>

Lecture note 1 and codes used in assignments