

# data\_exploration

January 26, 2017

## 0.1 loading data/libs

```
In [1]: import pandas as pd
import calendar
from bokeh.charts import output_notebook, Scatter, Bar, show, output_file, Line, BoxPlot
from bokeh.plotting import figure
from bokeh.io import hplot
output_notebook()
```

```
In [2]: INPUT="data/device_failure.csv"
dataset = pd.read_csv(INPUT, index_col=[0,1], parse_dates=[0])

label_dset = dataset[["failure"]]
```

### 0.1.1 checking devices

```
In [3]: total_failures_per_device = label_dset.groupby(level=1).agg(sum)
total_failures_per_device["failure"].value_counts()
```

```
Out[3]: 0    1062
        1     106
        Name: failure, dtype: int64
```

Each device fail at least once  
~10% device failing  
'only' 106 positive points

### 0.1.2 checking Dates

```
In [4]: dates = label_dset.index.get_level_values(0)
print "Range: from %s to %s" % (dates.min(), dates.max())

total_failures_per_date = label_dset.groupby(level=0).agg(sum)
print
print " n failures per date"
print str(total_failures_per_date["failure"].value_counts())
print
```

```

print "total: %i failures for %i days" % (total_failures_per_date["failure"].sum(),
total_failures_per_date[total_failures_per_date["failure"].sum()])

```

Range: from 2015-01-01 00:00:00 to 2015-11-02 00:00:00

```

n failures per date
0    228
1     54
2     19
3      2
8      1
Name: failure, dtype: int64

```

total: 106 failures for 76 days

```

In [5]: from bokeh.plotting import figure
data = total_failures_per_date.resample("M").sum()
test = label_dset.reset_index("device").resample("M").agg(lambda d : d.nunique())
data["n_devices"] = test["device"]
data["failure_ratio_percent"] = data["failure"] / data["n_devices"] * 100
data.index = (calendar.month_abbr[i] for i in data.index.month)
l = Line(
    data["failure_ratio_percent"],
    title="failures per Month",
    ylabel="% failure",
    xlabel="month"
)
show(l)

```

```

In [6]: l = Line(
    data["n_devices"],
    title="n devices seen per Month",
    ylabel="n_devices",
    xlabel="month"
)
show(l)

```

```

In [7]: weekday_dset = total_failures_per_date.copy()
weekday_dset.index = ["%i:%s" % (i,calendar.day_name[i]) for i in total_failures_per_date.index]

per_day = weekday_dset.groupby(level=0).sum()

print "failures per weekday"

per_day.sort_index()

```

failures per weekday

```
Out[7]:
```

	failure
0:Monday	27
1:Tuesday	18
2:Wednesday	15
3:Thursday	22
4:Friday	12
5:Saturday	8
6:Sunday	4

```
In [8]: from tabulate import tabulate
        # uncomment to print "markdown-compatible" output
        #d = per_day.sort_index()
        #print tabulate(d, headers = ["weekday", "NB failures" ],tablefmt="pipe")
```

- Long term trend with more failures in the past
- Less failures over the weekend
- The absence of weekend could be explained by maintenance hapening only during work-week (hence explaing more failures on monday

### 0.1.3 Per Device description

```
In [9]: import numpy as np
        dates = label_dset.swaplevel().reset_index("date")
        dd= dates["date"]
        devices = pd.DataFrame({"min_date":dd.groupby(level=0).min(),"failure":dates["failure"]})
        devices["max_date"] = dd.groupby(level=0).max()
        devices["n_lines"] = dd.groupby(level=0).count()
        devices["n_days"] = (devices["max_date"] - devices["min_date"] ) /np.timedelta64(1, 'D')
        devices["missing_values"] = devices["n_days"] - devices["n_lines"]
```

```
In [10]: devices["min_date"].value_counts()
```

```
Out[10]: 2015-01-01    1163
         2015-05-06         4
         2015-01-27         1
         Name: min_date, dtype: int64
```

### 0.1.4 checking the nb devices per month. this is better done above

```
In [11]: #pd.DataFrame({"n_devices":devices["max_date"].dt.month.value_counts().sort_index()})
        monthly_devices = pd.DataFrame({"n_devices":devices["max_date"].dt.month.value_counts()})
        monthly_devices.index = [calendar.month_abbr[i] for i in monthly_devices.index]
        monthly_devices
```

```
Out[11]:
```

	n_devices
Jan	399
Feb	46
Mar	184
Apr	112

May	72
Jun	6
Jul	15
Aug	150
Sep	38
Oct	115
Nov	31

### 0.1.5 bucketing the n devices with missing day data

```
In [12]: i = ( (devices["missing_values"] //20)*20).value_counts()
#i = ( (devices["missing_values"])).value_counts(bins=10)
i.index.name = "n missing days"
pd.DataFrame({"n devices":i.sort_index()})
```

```
Out[12]:
```

n missing days	n devices
-0.0	1077
20.0	26
40.0	21
60.0	8
80.0	3
100.0	28
120.0	4
140.0	1

```
In [13]: i = devices["n_days"].value_counts(bins=10).sort_index()
i.index.name='n_days'
b = Bar(pd.DataFrame(
    {"n_devices":i}),
    xlabel="n days",
    title="devices distributed by ndays"
)
show(b)
```

```
In [14]: failing_devices = devices[devices["failure"]>0].index
failing_devices_t = pd.DataFrame({"failure":label_dset["failure"].unstack().filter(items=failing_devices)})
def max_date(date):
    return np.max(date)

def failing_date(date):
    data = withdate.ix[date.index]
    return data[data["failure"]>0]["date"][0]

withdate = failing_devices_t.reset_index(level=1)
max_vs_failingdates = withdate.groupby(level=0).agg( {"date": [ max_date, failing_date] })
max_vs_failingdates.columns = max_vs_failingdates.columns.droplevel()
max_vs_failingdates["td"] = (max_vs_failingdates["max_date"] - max_vs_failingdates["failing_date"])/24
print
```

```

print "dt in days between first failure and end of measurement :"
print max_vs_failingdates["td"].value_counts()
print
print "n failures"
print max_vs_failingdates["sum"].value_counts()

```

dt in days between first failure and end of measurement :

0.0      101

2.0      2

30.0     1

1.0      1

12.0     1

Name: td, dtype: int64

n failures

1.0      106

Name: sum, dtype: int64

In [15]: print "looking at weird failures"

```

weird_devices = max_vs_failingdates[max_vs_failingdates["td"] > 0]

```

```

weirdos = failing_devices_t.reset_index(level=1).ix[set(weird_devices.index)]

```

```

print weirdos.set_index("date", append=True).unstack(level="device").to_string()

```

looking at weird failures

	failure				
device	S1F0GPFZ	S1F136J0	W1F0KCP2	W1F0M35B	W1F11ZG9
date					
2015-01-01	0.0	0.0	0.0	0.0	0.0
2015-01-02	0.0	0.0	0.0	0.0	0.0
2015-01-03	0.0	0.0	0.0	0.0	0.0
2015-01-04	0.0	0.0	0.0	0.0	0.0
2015-01-05	0.0	0.0	0.0	0.0	0.0
2015-01-06	0.0	0.0	0.0	0.0	0.0
2015-01-07	0.0	0.0	0.0	0.0	0.0
2015-01-08	0.0	0.0	0.0	0.0	0.0
2015-01-09	0.0	0.0	0.0	0.0	0.0
2015-01-10	0.0	0.0	0.0	0.0	0.0
2015-01-11	0.0	0.0	0.0	0.0	0.0
2015-01-12	0.0	0.0	0.0	0.0	0.0
2015-01-13	0.0	0.0	0.0	0.0	0.0
2015-01-14	0.0	0.0	0.0	0.0	0.0
2015-01-15	0.0	0.0	0.0	0.0	0.0
2015-01-16	0.0	0.0	0.0	0.0	0.0
2015-01-17	0.0	0.0	0.0	0.0	0.0
2015-01-18	0.0	0.0	0.0	0.0	0.0
2015-01-19	0.0	0.0	0.0	0.0	0.0

2015-01-20	0.0	0.0	0.0	0.0	0.0
2015-01-21	0.0	0.0	0.0	0.0	0.0
2015-01-22	0.0	0.0	0.0	0.0	0.0
2015-01-23	0.0	0.0	0.0	0.0	0.0
2015-01-24	0.0	0.0	0.0	0.0	0.0
2015-01-25	0.0	0.0	0.0	0.0	0.0
2015-01-26	0.0	0.0	0.0	0.0	0.0
2015-01-27	0.0	0.0	0.0	0.0	0.0
2015-01-28	0.0	0.0	0.0	0.0	0.0
2015-01-29	0.0	0.0	0.0	0.0	0.0
2015-01-30	0.0	0.0	0.0	0.0	0.0
2015-01-31	0.0	0.0	0.0	0.0	0.0
2015-02-01	0.0	0.0	0.0	0.0	0.0
2015-02-02	0.0	0.0	0.0	0.0	0.0
2015-02-03	0.0	0.0	0.0	0.0	0.0
2015-02-04	0.0	0.0	0.0	0.0	0.0
2015-02-05	0.0	0.0	0.0	0.0	0.0
2015-02-06	0.0	0.0	0.0	0.0	0.0
2015-02-07	0.0	0.0	0.0	0.0	0.0
2015-02-08	0.0	0.0	0.0	0.0	0.0
2015-02-09	0.0	0.0	0.0	0.0	0.0
2015-02-10	0.0	0.0	0.0	0.0	0.0
2015-02-11	0.0	0.0	0.0	0.0	0.0
2015-02-12	0.0	0.0	0.0	0.0	0.0
2015-02-13	0.0	0.0	0.0	0.0	0.0
2015-02-14	0.0	0.0	0.0	0.0	0.0
2015-02-15	0.0	0.0	0.0	0.0	0.0
2015-02-16	0.0	0.0	0.0	0.0	0.0
2015-02-17	0.0	0.0	0.0	0.0	0.0
2015-02-18	0.0	0.0	0.0	0.0	0.0
2015-02-19	0.0	0.0	0.0	0.0	0.0
2015-02-20	0.0	0.0	0.0	0.0	0.0
2015-02-21	0.0	0.0	0.0	0.0	0.0
2015-02-22	0.0	0.0	0.0	0.0	0.0
2015-02-23	0.0	0.0	0.0	0.0	0.0
2015-02-24	0.0	0.0	0.0	0.0	0.0
2015-02-25	0.0	0.0	0.0	0.0	0.0
2015-02-26	0.0	0.0	0.0	0.0	0.0
2015-02-27	0.0	0.0	0.0	0.0	0.0
2015-02-28	0.0	0.0	0.0	0.0	0.0
2015-03-01	0.0	0.0	0.0	0.0	0.0
2015-03-02	0.0	0.0	0.0	0.0	0.0
2015-03-03	0.0	0.0	0.0	0.0	0.0
2015-03-04	0.0	0.0	0.0	0.0	0.0
2015-03-05	0.0	0.0	0.0	0.0	0.0
2015-03-06	0.0	0.0	0.0	0.0	0.0
2015-03-07	0.0	0.0	0.0	0.0	0.0
2015-03-08	0.0	0.0	0.0	0.0	0.0

2015-03-09	0.0	0.0	0.0	0.0	0.0
2015-03-10	0.0	0.0	0.0	0.0	0.0
2015-03-11	0.0	0.0	0.0	0.0	0.0
2015-03-12	0.0	0.0	0.0	0.0	0.0
2015-03-13	0.0	0.0	0.0	0.0	0.0
2015-03-14	0.0	0.0	0.0	0.0	0.0
2015-03-15	0.0	0.0	0.0	0.0	0.0
2015-03-16	0.0	0.0	0.0	0.0	0.0
2015-03-17	0.0	0.0	0.0	0.0	0.0
2015-03-18	0.0	0.0	0.0	0.0	0.0
2015-03-19	0.0	0.0	0.0	0.0	0.0
2015-03-20	0.0	0.0	0.0	0.0	0.0
2015-03-21	0.0	0.0	0.0	0.0	0.0
2015-03-22	0.0	0.0	0.0	0.0	0.0
2015-03-23	0.0	0.0	0.0	0.0	0.0
2015-03-24	0.0	0.0	0.0	0.0	0.0
2015-03-25	0.0	0.0	0.0	0.0	0.0
2015-03-26	0.0	0.0	0.0	0.0	0.0
2015-03-27	0.0	0.0	0.0	0.0	0.0
2015-03-28	0.0	0.0	0.0	0.0	0.0
2015-03-29	0.0	0.0	0.0	0.0	0.0
2015-03-30	0.0	0.0	0.0	0.0	0.0
2015-03-31	0.0	0.0	0.0	0.0	0.0
2015-04-01	0.0	0.0	0.0	0.0	0.0
2015-04-02	0.0	0.0	0.0	0.0	0.0
2015-04-03	0.0	0.0	0.0	0.0	0.0
2015-04-04	0.0	0.0	0.0	0.0	0.0
2015-04-05	0.0	0.0	0.0	0.0	0.0
2015-04-06	0.0	0.0	0.0	0.0	0.0
2015-04-07	0.0	0.0	0.0	0.0	0.0
2015-04-08	0.0	0.0	0.0	0.0	0.0
2015-04-09	0.0	0.0	0.0	0.0	0.0
2015-04-10	0.0	0.0	0.0	0.0	0.0
2015-04-11	0.0	0.0	0.0	0.0	0.0
2015-04-12	0.0	0.0	0.0	0.0	0.0
2015-04-13	0.0	0.0	0.0	0.0	0.0
2015-04-14	0.0	0.0	0.0	0.0	0.0
2015-04-15	0.0	0.0	0.0	0.0	0.0
2015-04-16	0.0	0.0	0.0	0.0	0.0
2015-04-17	0.0	0.0	0.0	0.0	0.0
2015-04-18	0.0	0.0	0.0	0.0	0.0
2015-04-19	0.0	0.0	0.0	0.0	0.0
2015-04-20	0.0	0.0	0.0	0.0	0.0
2015-04-21	0.0	0.0	0.0	0.0	0.0
2015-04-22	0.0	0.0	0.0	0.0	0.0
2015-04-23	0.0	0.0	0.0	0.0	0.0
2015-04-24	0.0	0.0	0.0	0.0	0.0
2015-04-25	0.0	0.0	0.0	0.0	0.0

2015-04-26	0.0	0.0	0.0	0.0	0.0
2015-04-27	0.0	0.0	0.0	0.0	0.0
2015-04-28	0.0	0.0	0.0	0.0	0.0
2015-04-29	0.0	0.0	0.0	0.0	0.0
2015-04-30	0.0	0.0	0.0	0.0	0.0
2015-05-01	0.0	0.0	0.0	0.0	0.0
2015-05-02	0.0	0.0	0.0	0.0	0.0
2015-05-03	0.0	0.0	0.0	0.0	0.0
2015-05-04	0.0	0.0	0.0	0.0	0.0
2015-05-05	0.0	1.0	0.0	0.0	0.0
2015-05-06	0.0	0.0	0.0	0.0	0.0
2015-05-07	0.0	NaN	0.0	0.0	0.0
2015-05-08	0.0	NaN	0.0	0.0	0.0
2015-05-09	0.0	NaN	1.0	1.0	0.0
2015-05-10	0.0	NaN	0.0	0.0	0.0
2015-05-11	0.0	NaN	0.0	0.0	0.0
2015-05-12	0.0	NaN	NaN	NaN	0.0
2015-05-13	0.0	NaN	NaN	NaN	0.0
2015-05-14	0.0	NaN	NaN	NaN	0.0
2015-05-15	0.0	NaN	NaN	NaN	0.0
2015-05-16	0.0	NaN	NaN	NaN	0.0
2015-05-17	0.0	NaN	NaN	NaN	0.0
2015-05-18	0.0	NaN	NaN	NaN	0.0
2015-05-19	0.0	NaN	NaN	NaN	0.0
2015-05-20	0.0	NaN	NaN	NaN	0.0
2015-05-21	0.0	NaN	NaN	NaN	0.0
2015-05-22	0.0	NaN	NaN	NaN	0.0
2015-05-23	0.0	NaN	NaN	NaN	0.0
2015-05-24	0.0	NaN	NaN	NaN	0.0
2015-05-25	0.0	NaN	NaN	NaN	0.0
2015-05-26	0.0	NaN	NaN	NaN	0.0
2015-05-27	0.0	NaN	NaN	NaN	0.0
2015-05-28	0.0	NaN	NaN	NaN	0.0
2015-05-29	0.0	NaN	NaN	NaN	0.0
2015-05-30	0.0	NaN	NaN	NaN	0.0
2015-05-31	0.0	NaN	NaN	NaN	0.0
2015-06-01	0.0	NaN	NaN	NaN	0.0
2015-06-02	0.0	NaN	NaN	NaN	0.0
2015-06-03	0.0	NaN	NaN	NaN	0.0
2015-06-04	0.0	NaN	NaN	NaN	0.0
2015-06-05	0.0	NaN	NaN	NaN	0.0
2015-06-06	0.0	NaN	NaN	NaN	0.0
2015-06-07	0.0	NaN	NaN	NaN	0.0
2015-06-08	0.0	NaN	NaN	NaN	0.0
2015-06-09	0.0	NaN	NaN	NaN	0.0
2015-06-10	0.0	NaN	NaN	NaN	0.0
2015-06-11	0.0	NaN	NaN	NaN	0.0
2015-06-12	0.0	NaN	NaN	NaN	0.0



2015-06-13	0.0	NaN	NaN	NaN	0.0
2015-06-14	0.0	NaN	NaN	NaN	0.0
2015-06-15	0.0	NaN	NaN	NaN	0.0
2015-06-16	0.0	NaN	NaN	NaN	0.0
2015-06-17	0.0	NaN	NaN	NaN	0.0
2015-06-18	0.0	NaN	NaN	NaN	0.0
2015-06-19	0.0	NaN	NaN	NaN	0.0
2015-06-20	0.0	NaN	NaN	NaN	0.0
2015-06-21	0.0	NaN	NaN	NaN	0.0
2015-06-22	0.0	NaN	NaN	NaN	0.0
2015-06-23	0.0	NaN	NaN	NaN	0.0
2015-06-24	0.0	NaN	NaN	NaN	0.0
2015-06-25	0.0	NaN	NaN	NaN	0.0
2015-06-26	0.0	NaN	NaN	NaN	0.0
2015-06-27	0.0	NaN	NaN	NaN	0.0
2015-06-28	0.0	NaN	NaN	NaN	0.0
2015-06-29	0.0	NaN	NaN	NaN	0.0
2015-06-30	0.0	NaN	NaN	NaN	0.0
2015-07-01	0.0	NaN	NaN	NaN	0.0
2015-07-02	0.0	NaN	NaN	NaN	0.0
2015-07-03	0.0	NaN	NaN	NaN	0.0
2015-07-04	0.0	NaN	NaN	NaN	0.0
2015-07-05	0.0	NaN	NaN	NaN	0.0
2015-07-06	0.0	NaN	NaN	NaN	0.0
2015-07-07	0.0	NaN	NaN	NaN	0.0
2015-07-08	0.0	NaN	NaN	NaN	0.0
2015-07-09	0.0	NaN	NaN	NaN	0.0
2015-07-10	0.0	NaN	NaN	NaN	0.0
2015-07-11	0.0	NaN	NaN	NaN	0.0
2015-07-12	1.0	NaN	NaN	NaN	0.0
2015-07-13	0.0	NaN	NaN	NaN	0.0
2015-07-14	0.0	NaN	NaN	NaN	0.0
2015-07-15	0.0	NaN	NaN	NaN	0.0
2015-07-16	0.0	NaN	NaN	NaN	0.0
2015-07-17	0.0	NaN	NaN	NaN	0.0
2015-07-18	0.0	NaN	NaN	NaN	1.0
2015-07-19	0.0	NaN	NaN	NaN	0.0
2015-07-20	0.0	NaN	NaN	NaN	0.0
2015-07-21	0.0	NaN	NaN	NaN	0.0
2015-07-22	0.0	NaN	NaN	NaN	0.0
2015-07-23	0.0	NaN	NaN	NaN	0.0
2015-07-24	0.0	NaN	NaN	NaN	0.0
2015-07-25	NaN	NaN	NaN	NaN	0.0
2015-07-26	NaN	NaN	NaN	NaN	0.0
2015-07-27	NaN	NaN	NaN	NaN	0.0
2015-07-28	NaN	NaN	NaN	NaN	0.0
2015-07-29	NaN	NaN	NaN	NaN	0.0
2015-07-30	NaN	NaN	NaN	NaN	0.0

2015-07-31	NaN	NaN	NaN	NaN	0.0
2015-08-01	NaN	NaN	NaN	NaN	0.0
2015-08-02	NaN	NaN	NaN	NaN	0.0
2015-08-03	NaN	NaN	NaN	NaN	0.0
2015-08-04	NaN	NaN	NaN	NaN	0.0
2015-08-05	NaN	NaN	NaN	NaN	0.0
2015-08-06	NaN	NaN	NaN	NaN	0.0
2015-08-07	NaN	NaN	NaN	NaN	0.0
2015-08-08	NaN	NaN	NaN	NaN	0.0
2015-08-09	NaN	NaN	NaN	NaN	0.0
2015-08-10	NaN	NaN	NaN	NaN	0.0
2015-08-11	NaN	NaN	NaN	NaN	0.0
2015-08-12	NaN	NaN	NaN	NaN	0.0
2015-08-13	NaN	NaN	NaN	NaN	0.0
2015-08-14	NaN	NaN	NaN	NaN	0.0
2015-08-15	NaN	NaN	NaN	NaN	0.0
2015-08-16	NaN	NaN	NaN	NaN	0.0
2015-08-17	NaN	NaN	NaN	NaN	0.0

- identified a list of devices, which are still measured after having failed.

Three hypothesis: - The device is still fonctionnal after maintenance - The failure was a fluke - The measurement thereafter are false

==> if we cannot distinguish between these hypothesis, need to remove these devices from the dataset

In [ ]: