

Crunching Gigabytes Locally

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Data crunching TL;DR

- **The #1 productivity booster in data science is fast iterations.**
- Use the shell for under 10MB, use Hadoop for over 100GB.
- In between there's room for creativity, and it is what this talk is about.

About me

- Machine learning, data science, and feature engineering for over 10 years.
- Competitive programming, algorithms, infra, and C/C++ for over 20 years.



REALTYSHARES



About the measurements

- Timings measured on my laptop, which is:
 - ThinkPad X1 Carbon 4th gen.
 - On AC power and on battery, no significant difference.
 - Intel Core i7-7500U, 3.5 GHz, 4MB cache.
 - 16GB LPDDR3 1866.
 - 256 GB SSD, OPAL2.0 PCIe-NVMe.
- Also validated on a Hetzner box, with no surprises.
- All code snippets are [on GitHub](#).



About the dataset

- [2016 Green Taxi Trip Data](#) (hosted by Google for BigQuery, [link](#)).
- 2.1GB, 16.3M rides.
- Generally speaking, dataset integrity, as well as licensing, are real things.
- If you want the results reproducible, keep immutable copies of the datasets.



Data cleanup

- You would imagine the Google-hosted dataset is clean enough. Well, no.
- Two major sources of issues:
 - Timestamps.
 - Ultra-short and/or ultra-fast rides.

Issues with timestamps

- Ambiguity. Real example: 07/11/2016 01:05:09 AM

```
$ TZ=America/New_York date -d @1478408709  
Sun Nov 6 01:05:09 EDT 2016  
$ TZ=America/New_York date -d @1478412309  
Sun Nov 6 01:05:09 EST 2016
```

- Straight out invalid dates. Real example: 03/13/2016 02:58:40 AM

```
$ TZ=America/New_York date -d '03/13/2016 02:58:40 AM'  
date: invalid date '03/13/2016 02:58:40 AM'  
$ TZ=America/New_York date -d @$(( $(TZ=America/New_York date -d '03/13/2016 01:59:59' +%s) + 1))  
Sun= Mar 13 03:00:00 EDT 2016
```

- Trip into the past.

Example: 07/17/2016 03:16:35 PM, 07/16/2016 08:30:10 AM



Lesson learned

- Human time is hard to deal with for computers.
 - Unix epoch timestamps are the safest way.
 - If “derived” data, such as day of week, is needed, keep in mind it may change!
 - Even a dataset published by Google is not immune to this.
- **Solution:** Re-annotate timestamps in the input CSV.
 - For this one-time task, use Python to keep things simple.

Crunching tasks

- Follow the first two examples from the [dataset page](#) on Google cloud:
 - How many trips did Yellow taxis take each month in 2015?
 - What was the average speed of Yellow taxi trips in 2015?

Crunching

- How many trips did Yellow taxis take each month in 2015?

Crunching

- How many trips did Yellow taxis take each month in 2015?

```
$ ./step1_rides_by_months_python.py | tee >(md5sum)
```

```
1445285 01/2016
```

```
1510722 02/2016
```

```
1576393 03/2016
```

```
1543925 04/2016
```

```
1536979 05/2016
```

```
1404726 06/2016
```

```
1332510 07/2016
```

```
1247675 08/2016
```

```
1162373 09/2016
```

```
1252572 10/2016
```

```
1148214 11/2016
```

```
1224158 12/2016
```

```
bd5b6d7f389e64aab50c56fed61eaded -
```

Python timing

- The Python script runs for 24 seconds.
- Take a moment to think of this number.

Is 24 seconds fast or slow?

History

- Python is a fine scripting language, but what did people use before it?

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- Aho-Weinberger-Kernighan.
 - Or [awk](#).
 - From the 70s.
 - One can say it has a built-in map-reduce.
 - And, at 16 seconds running time, it is actually faster.
- Shell classics: grep, cut, sort, uniq.
 - This approach is even faster, completing in 12 seconds total.
 - 2x the Python speed.
 - Truly, “without any programming”.

Shell

- What we should really be asking ourselves is:
 - How long does it take to look for something in the input data, such as that bad timestamp?
 - Wait, and how long does it take to count the number of lines in the file?

Shell

- What we should really be asking ourselves is:
 - How long does it take to look for something in the input data, such as that bad timestamp?
 - Wait, and how long does it take to count the number of lines in the file?
- Both are valid questions, both require full file scan, both are doable in shell.

Timings:

- `grep ...` : ~ 0.9s (for what's only found a few times in the input)
 - `wc -l ...` : < 0.5s
- In other words, **grep** and **wc** operate at over 1GB/s throughput.

All timings

Approach	Duration
python	~24s (<100MB/s)
"very naive" c++	~18s (~27s on Hetzner)
awk	~17s
cut, sort, uniq	~14s
"naive" c++, string views instead of strings	~13s
read full file, malloc() + { open()+read() / fopen()+fread() }	~3s / ~2.4s
mmap() full file, C	1.4s (>1GB/s)
"naive" C, fopen() + fgets(), analyze digits at fixed string offset	1.0s

Python, at ~24s

```
histogram = {}
```

```
with open('../cooked.csv') as f:
```

```
    for line in f:
```

```
        date = line.split(',')[1]
```

```
        yyyy_mm_dd = date.split('-')
```

```
        yyyy = yyyy_mm_dd[0]
```

```
        mm = yyyy_mm_dd[1]
```

```
        histogram_key = mm + '/' + yyyy
```

```
        if not histogram_key in histogram:
```

```
            histogram[histogram_key] = 1
```

```
        else:
```

```
            histogram[histogram_key] += 1
```

```
for mm_yyyy, count in sorted(histogram.items()):
```

```
    print "%s %s" % (count, mm_yyyy)
```



awk, at ~17s

Keep in mind, this code is from the 70s!

```
BEGIN { FS="," }
```

```
{  
    split($2, date_time, " ");  
    split(date_time[1], yyyy_mm_dd, "-");  
    ++histogram[yyyy_mm_dd[2] "/" yyyy_mm_dd[1]];  
}
```

```
END { for (i in histogram) printf "%d %s\n", histogram[i], i; }
```

The winner, plain C, consistently at 1.0s

```
#include <stdio.h>

int MM[12];

int main(int argc, char** argv) {
    char line[10000]; // Unsafe, of course.

    FILE* f = fopen(argc >= 2 ? argv[1] : "../cooked.csv", "r");
    while (fgets(line, sizeof(line), f)) {
        ++MM[(line[7] - '0') * 10 + (line[8] - '0') - 1]; // Unsafe, of course.
    }
    fclose(f);

    for (int m = 0; m < 12; ++m) {
        printf("%d %02d/%d\n", MM[m], m + 1, 2016);
    }
}
```

Lessons learned

- Unix rocks.
 - When it comes to textual throughput, shell-grade tools are the fastest.
 - Unsophisticated old-school [**unsafe**] plain C is not far behind.

Lessons learned

- Unix rocks.
 - When it comes to textual throughput, shell-grade tools are the fastest.
 - Unsophisticated old-school [**unsafe**] plain C is not far behind.
- Scripting is ~20x slower.
 - Even splitting the row into vector<string> increases the running time from 13s to 18s.
 - In other words, just creating temporary strings each row takes ~33% of CPU cycles.
 - Or, to put it another way, meanwhile, the “**dumb C**” solution could have run five times.
 - Here goes a rant about using a “high-level language” such as Java for storage-heavy tasks.
- Python can be read by more people, but it's a pain for the CPUs.

Going native

- The fastest code snippets are less readable, as CSV is not natural for C.
- To achieve higher speeds, machine-friendly data formats should be used.

```
void Run(const struct Ride* data, size_t n) {  
    for (size_t i = 0; i < n; ++i) {  
        ++MM[data[i].pickup.month - 1]  
    }  
    for (int m = 0; m < 12; ++m) {  
        printf("%d %02d/2016\n", MM[m], m + 1);  
    }  
}
```

- Runs in under 0.2s, including mmap()-ing the entire 1.5GB binary input file.

Crunching, exercise two

- What was the average speed of Yellow taxi trips in 2015?

Crunching, exercise two

- What was the average speed of Yellow taxi trips in 2015?

```
$ gcc -O3 step3_sanitycheck_avg_speed_by_hour.c && time ./a.out | tee >(md5sum)
```

```
Total rides considered: 15592085 (95.2%)
```

```
00 15.37
01 15.85
02 16.10
03 16.67
04 17.81
05 19.72
06 19.47
07 14.82
08 12.99
09 13.55
10 13.49
11 13.51
12 13.22
13 13.18
14 12.40
15 12.02
16 11.61
17 11.29
18 11.74
19 12.70
20 13.51
21 14.00
22 14.46
23 14.93
```

Analysis

- Rides filtering logic blindly copied from the BigQuery example.
 - And the result is reasonable compared to the BigQuery example.
- The computation uses the same constraints as the original code snippet.
 - Including **“cost per mile is between \$2 and \$10”**.
 - Which is **floating point**.

Non-integers

- Comparison and rounding are always a path to trouble.
- Especially if we want results that are reproducible in various ways.
- Generally, if floating point can be avoided, avoid it.

Floating point surprises: three (!) different results

```
1 #include "schema.h"
2 void Run(const struct Ride* data, int n) {
3     int total_considered = 0;
4     #if 0
5         const int two = 2;
6         const int ten = 10; // `float` and `double` give different results here. -- D.K.
7     #endif
8     for (int i = 0; i < n; ++i) {
9         const int trip_duration_seconds = data[i].dropoff.epoch - data[i].pickup.epoch;
10        if (data[i].trip_distance > 0 && trip_duration_seconds > 0) {
11            #if 1
12                const double cost_per_mile = data[i].fare_amount / data[i].trip_distance;
13                if (cost_per_mile >= 2.0 && cost_per_mile <= 10.0) {
14            #else
15                if (data[i].fare_amount >= two * data[i].trip_distance &&
16                    data[i].fare_amount <= ten * data[i].trip_distance) {
17            #endif
18                ++total_considered;
19                // UpdateAverageSpeed(data[i].pickup.hour,
20                //                      data[i].trip_distance / trip_duration_seconds);
21            }
22        }
23    }
24 }
```

Crunching, do over

- What was the average speed of Yellow taxi trips in 2015?

```
$ gcc -O3 step3_sanitycheck_avg_speed_by_dow.c && time ./a.out | tee >(md5sum)
```

```
Total rides considered: 15592077 (95.2%)
```

```
00 15.37
01 15.85
02 16.10
03 16.67
04 17.81
05 19.72
06 19.47
07 14.82
08 12.99
09 13.55
10 13.49
11 13.51
12 13.22
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15 12.02
16 11.61
17 11.29
18 11.74
19 12.70
20 13.51
21 14.00
22 14.46
23 14.93
```

Dealing with non-integers

- In the yellow cabs case:
 - All times are in seconds, which are integers.
 - All amounts are in dollars, which are integer cents.
 - Distance can be rounded, and let's round to 1/100th of a mile to simplify division.
- **Solution:** Update the schema to use integers, and re-cook the binary file.
 - The file can even get smaller thanks to it.

But wait, there's more!

```
for (size_t i = 0; i < n; ++i) {
    const int trip_duration_s = data[i].dropoff.epoch - data[i].pickup.epoch;
    if (data[i].trip_distance_times_100 > 0 && trip_duration_s > 0) {
        if (data[i].fare_amount_cents >= 2 * data[i].trip_distance_times_100 &&
            data[i].fare_amount_cents <= 10 * data[i].trip_distance_times_100) {
            ++total_considered;
            UpdateAverageSpeed(&per_hour_counters[data[i].pickup.hour],
                              data[i].trip_distance_times_100,
                              trip_duration_seconds);
        }
    }
}
```

// The true average cab speed is total miles traveled divided by total hours traveled.
// The average cab speed and the average of average ride speeds are not the same!
// The original code computed something else, and rewriting the code helped locate it.

Recap

- A.k.a. “so what?”:
 - ***We know Python is slow, but developers understand it.***
Their time is still more valuable than the time the machine spends running their code.
Thanks, we’re not considering sed or awk.
 - ***We know mmap()-ing a file is fast, thanks, Captain Obvious.***
We don’t plan on doing it though, sorry.
 - ***We know binary formats are fast.***
Would be nice had they been some Protobuf or Thrift or Cap’n Proto or Avro.
For pure binary, we, of course, have no plans to switch to C/C++.

The real thing

- Move code to data, not data to code.
 - In this particular example we `mmap()`-ed the file.
 - Which is OK for a few GBs, but not fun for a file that barely fits in memory.
 - Still keeping a weak laptop in mind.

The real thing

- Move code to data, not data to code.
 - In this particular example we `mmap()`-ed the file.
 - Which is OK for a few GBs, but not fun for a file that barely fits in memory.
 - Still keeping a weak laptop in mind.
- We need another superpower to complement `mmap()`, and it's `dlopen()`.
 - Load a library dynamically, and natively call user-defined functions from it.

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- Move code to data, not data to code.
 - In this particular example we `mmap()`-ed the file.
 - Which is OK for a few GBs, but not fun for a file that barely fits in memory.
 - Still keeping a weak laptop in mind.
- We need another superpower to complement `mmap()`, and it's `dlopen()`.
 - Load a library dynamically, and natively call user-defined functions from it.
- The production design then is:
 - The binary, that has all the data loaded up, runs as a daemon.
 - Introduce a lightweight interface to run “custom queries” against it.

Worth a thousand words!

- It's in the browser.

Script

```
FUNCTION(rides, output) {  
    std::vector<int> MM(12);  
  
    for (const IntegerRide& ride : rides) {  
        ++MM[ride.pickup.month - 1];  
    }  
  
    for (int m = 0; m < 12; ++m) {  
        output << Printf("%d %02d/2016\n", MM[m], m + 1);  
    }  
}
```

Run

Endpoint

[Link](#)

Result

```
1445285 01/2016  
1510722 02/2016  
1576393 03/2016  
1543925 04/2016  
1536979 05/2016  
1404726 06/2016  
1332510 07/2016  
1247675 08/2016  
1162373 09/2016  
1252572 10/2016  
1148214 11/2016  
1224158 12/2016
```

How does it work

- An obligatory disclaimer: I am not a frontend developer.
 - The browser UX is just a demo, we never used it, really.
 - Obviously, syntax highlighting, completion, and on-the-fly compilation can be added.
- The user code is wrapped into a boilerplate for compilation.
 - The boilerplate is service-defined, hence `Printf()` and the `FUNCTION()` macro above.
 - The code is then SHA256-hashed, and each unique hash is only compiled once.
 - The user code can accept parameters, from the URL/querystring, or from HTTP headers.
- The compilation is run from a dedicated temporary directory.
 - With a symlink to the framework, to enable accessing extra headers as necessary.
 - With a dummy, empty, header file the user could `#include` locally to “fix” completion.
- A `#line` directive is injected between the boilerplate and the user code.
 - So that the IDE can jump to exact locations of the error should one occur in user code.

Also, completion! (No, this pink is not my color scheme. I code on a dark background.)

```
1 #include "current.h"
2
3 FUNCTION(rides, output) {
4     std::vector<int> MM(12);
5
6     for (auto const& ride : rides) {
7         if (ride.pickup.█
8             epoch          m int
9             for (int m = 0; m dow          m uint8_t
10                output << Print month          m uint8_t m], m + 1);
11        }                Timestamp:: s
12    }                year          m uint8_t
~                second          m uint8_t
~                dst          m uint8_t
~                day          m uint8_t
~                hour          m uint8_t
~                minute          m uint8_t
```

How does it work cont'd

- To function flawlessly, best is to put three more files alongside the “script”.
 - The dummy header file.
 - To define the same symbols as the boilerplate, such as `FUNCTION()`.
 - This enables proper syntax highlighting and completion.
 - In fact, in the demo it's a symlink to the boilerplate.
 - A symlink to the very framework.
 - So that “library” functions such as `Printf()` or `JSON()` are easy to access.
 - A Makefile.
 - To push the code to the service using `curl`.
 - To “build” the code and “run” it, if built successfully.
 - To set the HTTP header with the original source name, for the IDE to jump to errors.

Bonus: JSON support and auto-set Content-Type

```
$ cat snippet.txt
```

```
ENDPOINT(data, params) {  
    std::map<int, int> histogram;  
    for (const IntegerRide& ride : data) {  
        ++histogram[ride.pickup.month - 1];  
    }  
    return histogram;  
}
```

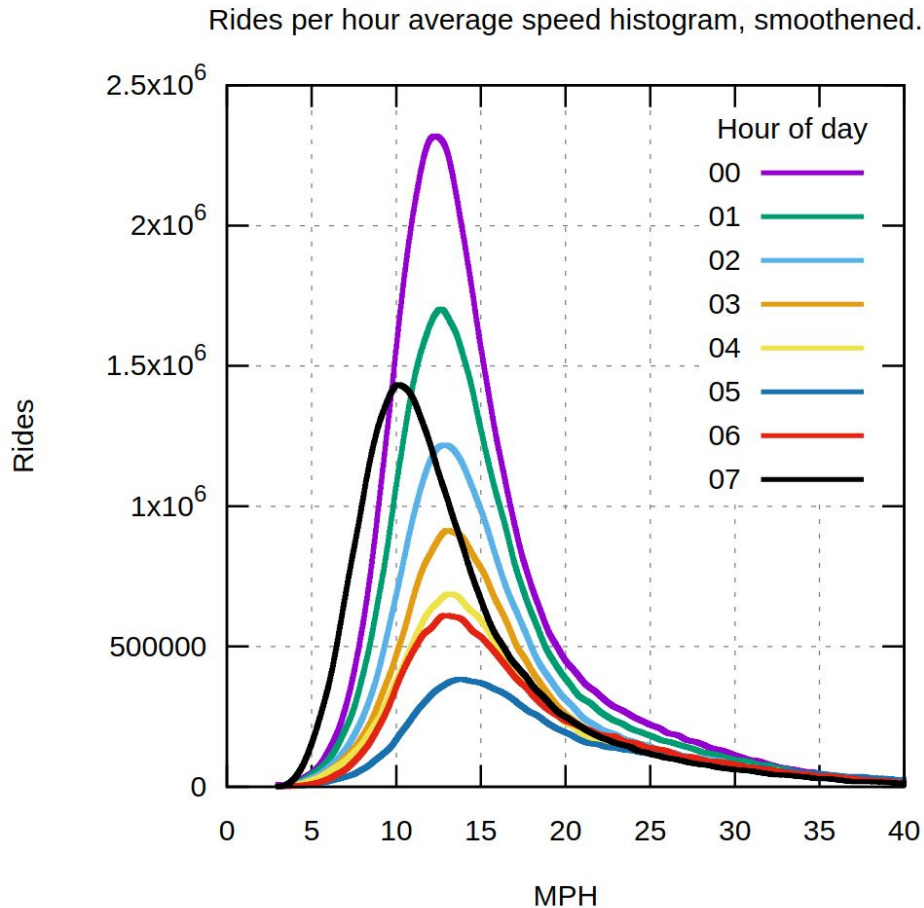
```
$ curl --data-binary @snippet.txt http://localhost:3000/full # Takes a bit longer, as JSON support is some wild metaprogramming. -- D.K.  
Compiled: @54108e84c0
```

```
$ curl -i http://localhost:3000/full/54108e84c0  
HTTP/1.1 200 OK  
Content-Type: application/json; charset=utf-8  
Connection: close  
Access-Control-Allow-Origin: *  
Content-Length: 148
```

```
[ [0,1445285], [1,1510722], [2,1576393], [3,1543925], [4,1536979], [5,1404726], [6,1332510], [7,1247675], [8,1162373], [9,1252572], [10,1148214], [11,1224158] ]
```


Bonus II: Visualizations

- **gnuplot**
 - Using the existing binding.
- **Compile once.**
 - Use URL parameters later.
 - `/e3fcdadf67?from=0&to=8`
- **Ready to use endpoint.**
 - For monitoring and reporting.



Pros, aside from performance and zero configuration

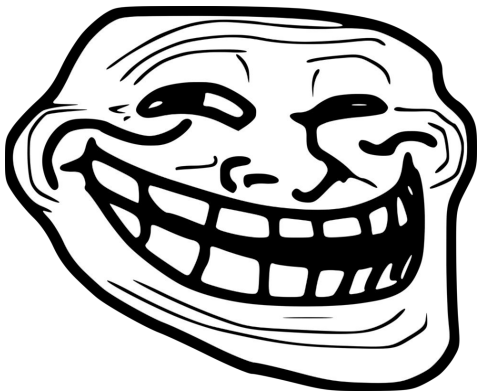
- Brings together the best of both worlds:
 - For the old-schoolers:
 - Code completion.
 - Build and run with one command, `Makefile`-friendly.
 - Every functional code snippet immediately becomes a `curl`-able endpoint.
 - For browser-cherishing folks:
 - Zero entry barrier.
 - Even this trivial demo is sharing-enabled.
 - User-defined logic can accept URL parameters.
- If the backend is in C/C++, the above is a prototyping framework.
 - Production-ready for internal alerts, monitoring, and reporting.
 - A playground for real-time ML features.
 - The very same code then goes through the review, and makes it as a prod API endpoint.

Cons

- Zero security.
 - Bad memory access would SEGFAULT the serving daemon.
 - I'm not sharing this very URL, as one could run `rm -rf /` on my machine.
 - Can run from under an unprivileged user.
 - Not really an issue with modern virtualization and containers.
 - Even less of an issue as data scientists tend to self-organize into friendly sub-teams.
- You have to write C++.
 - I made it as “JavaScript” as possible for you, but, face it, it is C++ behind the scenes.
 - Although one can argue whether it's a bad thing.

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Conclusions and Q&A

- “Big” data is in the eye of the *shell holder*.
 - For single- and even double-digit gigabytes, most crunching still fits a laptop.
- Streaming data in real time is also no problem whatsoever.
 - As long as the daemon binary is listening to some pub-sub bus, which most of us have.
- Don’t rush to use fancy trending toolkits until they are needed.
 - I personally have stopped people from using tools for the sake of using them.
- First principles and C++ rock.
 - And C++ is a simple and safe enough language by 2018.
If you speak JS, you can do modern C++, and you will enjoy it.

Thanks

Q&A