# Using Dask DataFrames

PARALLEL COMPUTING WITH DASK



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### Reading CSV

import dask.dataframe as dd

- dd.read\_csv() function
  - Accepts single filename or glob pattern (with wildcard \* )
  - Does not read file immediately (lazy evaluation)
  - File(s) need not fit in memory

#### Reading multiple CSV files

%ls

```
quarter1.csv quarter2.csv quarter3.csv quarter4.csv
```

```
transactions = dd.read_csv('*.csv')
```

```
transactions.head()
transactions.tail()
```

```
id
                            date
             amount
       names
              -1159
131
     Norbert
                     2016-01-01
              1149
                     2016-01-01
       Jerry
342
                     2016-01-01
485
        Dan
               1380
               1555 2016-01-02
     Xavier
513
    Michael
                    2016-01-02
849
                363
```

	id	names	amount	date
195	838	Wendy	87	2016-12-28
196	915	Bob	852	2016-12-30
197	749	Patricia	1741	2016-12-31
198	743	Michael	1191	2016-12-31
199	889	Wendy	336	2016-12-31

### Building delayed pipelines

```
is_wendy = (transactions['names'] == 'Wendy')
wendy_amounts = transactions.loc[is_wendy, 'amount']
wendy_amounts
```

```
Dask Series Structure:
npartitions=4
None int64
None ...
None ...
None ...
None ...
None ...
Name: amount, dtype: int64
Dask Name: loc-series, 24 tasks
```



#### Building delayed pipelines

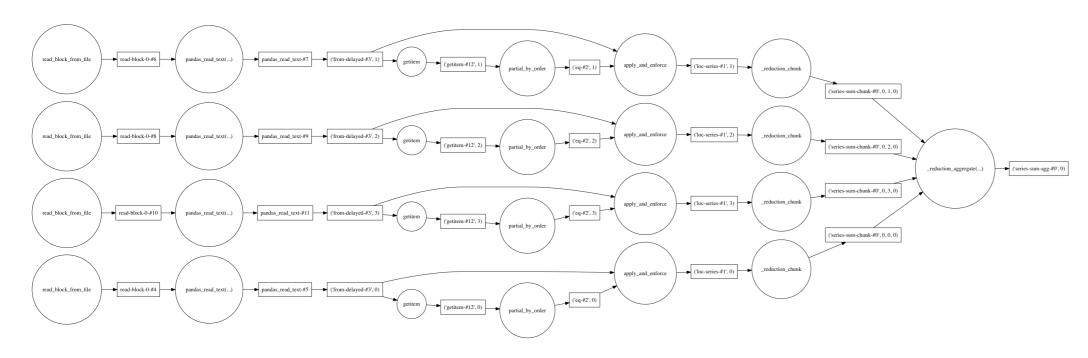
```
wendy_diff = wendy_amounts.sum()
wendy_diff
```

```
dd.Scalar<series-..., dtype=int64>
```

```
wendy_diff.visualize(rankdir='LR')
```



## Visualizing pipelines



#### Compatibility with Pandas API

#### Unavailable in dask.dataframe:

- some unsupported file formats (e.g., .xls , .zip , .gz )
- sorting

#### Available in dask.dataframe:

- indexing, selection, & reindexing
- aggregations: .sum() , .mean() , .std() , .min() ,.max() etc.
- grouping with .groupby()
- datetime conversion with dd.to\_datetime()

## Let's practice!

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# Timing DataFrame Operations

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#### How big is big data?

DatasizeM	Required hardware	
$M < 8\mathrm{GB}$	RAM (single machine)	
$8\mathrm{GB} < M < 10\mathrm{TB}$	hard disk (single machine)	
$M>10\mathrm{TB}$ :	specialized hardware	

#### Two key questions:

- Data fits in RAM (random access memory)?
- Data fits on hard disk?

#### Taxi CSV files

```
%ll -h yellow_tripdata_2015-*.csv
```

```
1.8G 31 Jul 16:43 yellow_tripdata_2015-0
      staff
 user
              1.8G 31 Jul 16:43 yellow_tripdata_2015-0
       staff
 user
              1.9G 31 Jul 16:43 yellow_tripdata_2015-0
       staff
 user
              1.9G 31 Jul 16:43 yellow_tripdata_2015-0
       staff
 user
              staff
1 user
       staff
              1.8G 31 Jul 16:43 yellow_tripdata_2015-0
1 user
              staff
1 user
              1.6G 31 Jul 16:43 yellow_tripdata_2015-0
       staff
1 user
       staff
              1.6G 31 Jul 16:43 yellow_tripdata_2015-0
1 user
       staff
              1.8G 31 Jul 16:43 yellow_tripdata_2015-7
 user
              1.7G 31 Jul 16:43 yellow_tripdata_2015-7
       staff
1 user
              1.7G 31 Jul 16:43 yellow_tripdata_2015-1
      staff
1 user
```



### Timing I/O & computation: Pandas

```
import time, pandas as pd
t_start = time.time();
df = pd.read_csv('yellow_tripdata_2015-01.csv');
t_end = time.time();
print('pd.read_csv(): {} s'.format(t_end-t_start)) # time [s]
```

```
pd.read_csv: 43.820565938949585 s
```

```
t_start = time.time();
m = df['trip_distance'].mean();
t_end = time.time();
print('.mean(): {} ms'.format((t_end-t_start)*1000)) # time [ms]
```

```
.mean(): 17.752885818481445 ms
```

### Timing I/O & computation: Dask

```
import dask.dataframe as dd, time

t_start = time.time();

df = dd.read_csv('yellow_tripdata_2015-*.csv');

t_end = time.time();

print('dd.read_csv: {} ms'.format((t_end-t_start)*1000)) # time [ms
```

```
dd.read_csv: 404.7999382019043 ms
```

```
t_start = time.time();
m = df['trip_distance'].mean();
t_end = time.time();
print('.mean(): {} ms'.format((t_end-t_start)*1000)) # time [ms]
```

```
.mean(): 2.289295196533203 ms
```



### Timing I/O & computation: Dask

```
t_start = time.time();
result = m.compute();
t_end = time.time();
print('.compute(): {} min'.format((t_end-t_start)/60)) # time [min]
```

.compute(): 3.4004417498906454 min

### Timing in the IPython shell

```
m = df['trip_distance'].mean()
%time result = m.compute()
```

```
CPU times: user 9min 50s, sys: 1min 16s, total: 11min 7s
Wall time: 3min 1s
```

#### Is Dask or Pandas appropriate?

- How big is dataset?
- How much RAM available?
- How many threads/cores/CPUs available?
- Are Pandas computations/formats supported in Dask API?
- Is computation I/O-bound (disk-intensive) or CPU-bound (processor intensive)?

#### Best use case for Dask

- Computations from Pandas API available in Dask
- Problem size close to limits of RAM, fits on disk

## Let's practice!

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# Analyzing NYC Taxi Rides

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#### The New York taxi dataset





#### Taxi CSV files

```
%ll -h yellow_tripdata_2015-*.csv
```

```
1.8G 31 Jul 16:43 yellow_tripdata_2015-01.csv
          1 user staff
                           1.8G 31 Jul 16:43 yellow_tripdata_2015-02.csv
                  staff
           1 user
                           1.9G 31 Jul 16:43 yellow_tripdata_2015-03.csv
          1 user staff
                           1.9G 31 Jul 16:43 yellow_tripdata_2015-04.csv
          1 user staff
                           1.9G 31 Jul 16:43 yellow_tripdata_2015-05.csv
-rw-r--r-- 1 user staff
                           1.8G 31 Jul 16:43 yellow_tripdata_2015-06.csv
-rw-r--r-- 1 user staff
                           1.7G 31 Jul 16:43 yellow_tripdata_2015-07.csv
-rw-r--r-- 1 user staff
                           1.6G 31 Jul 16:43 yellow_tripdata_2015-08.csv
-rw-r--r-- 1 user staff
-rw-r--r-- 1 user staff
                           1.6G 31 Jul 16:43 yellow_tripdata_2015-09.csv
                           1.8G 31 Jul 16:43 yellow_tripdata_2015-10.csv
-rw-r--r-- 1 user staff
                           1.7G 31 Jul 16:43 yellow_tripdata_2015-11.csv
-rw-r--r-- 1 user staff
                           1.7G 31 Jul 16:43 yellow_tripdata_2015-12.csv
-rw-r--r-- 1 user staff
```

Exercises use smaller files...

#### Taxi data features

```
import pandas as pd

df = pd.read_csv('yellow_tripdata_2015-01.csv')

df.shape

df.columns
```



#### **Amount paid**

- How much was each ride?
  - o fare\_amount : cost of ride
  - o tolls\_amount : charges for toll roads
  - extra : additional charges
  - tip\_amount : amount tipped (credit cards only)
  - total\_amount : total amount paid by passenger



#### Payment type

```
df['payment_type'].value_counts()
```

```
1  7881388
2  4816992
3  38632
4  11972
5  2
Name: payment_type, dtype: int64
```



## Let's practice!

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