

Components of a data platform

BUILDING DATA ENGINEERING PIPELINES IN PYTHON

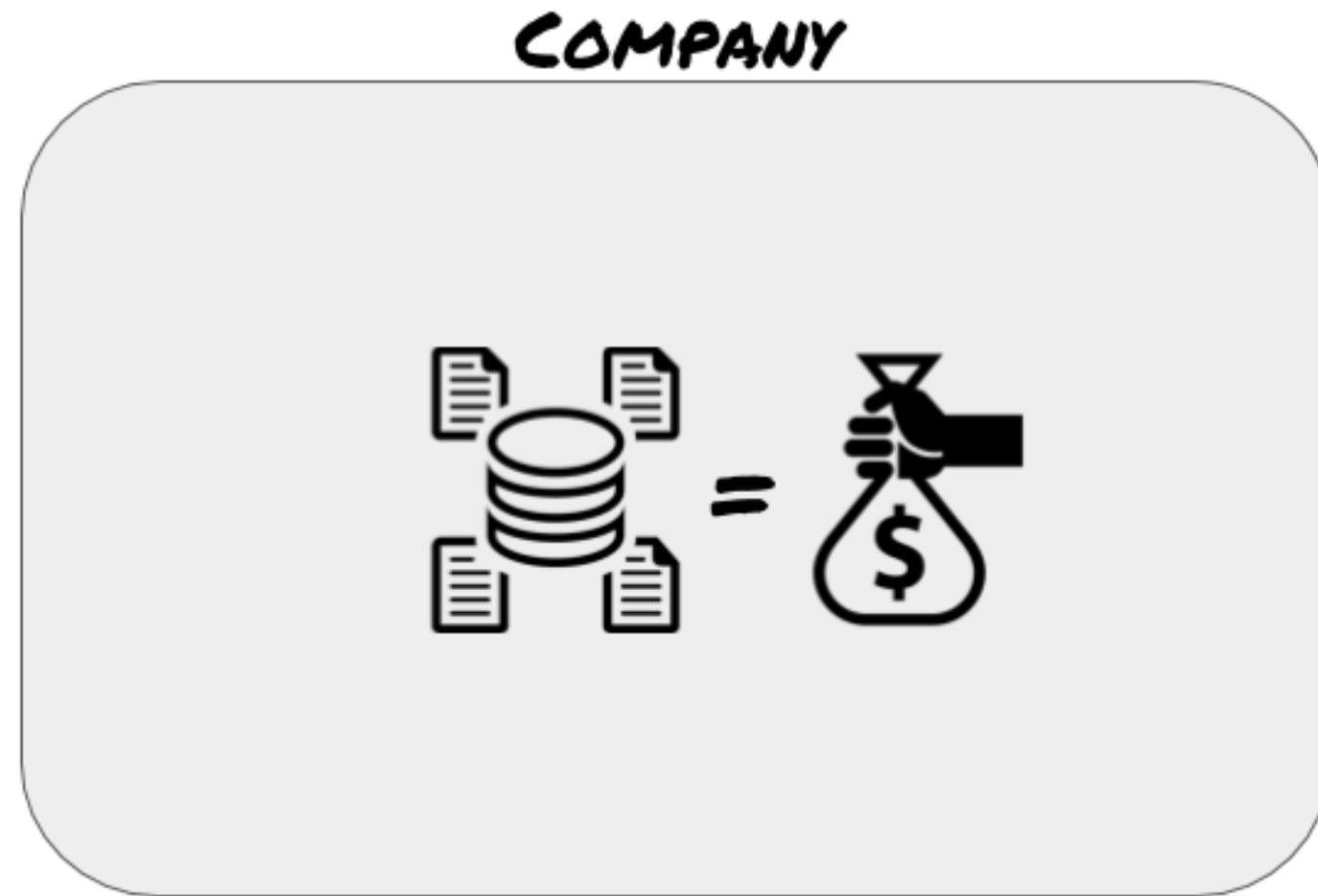


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Data Engineer at Data Minded

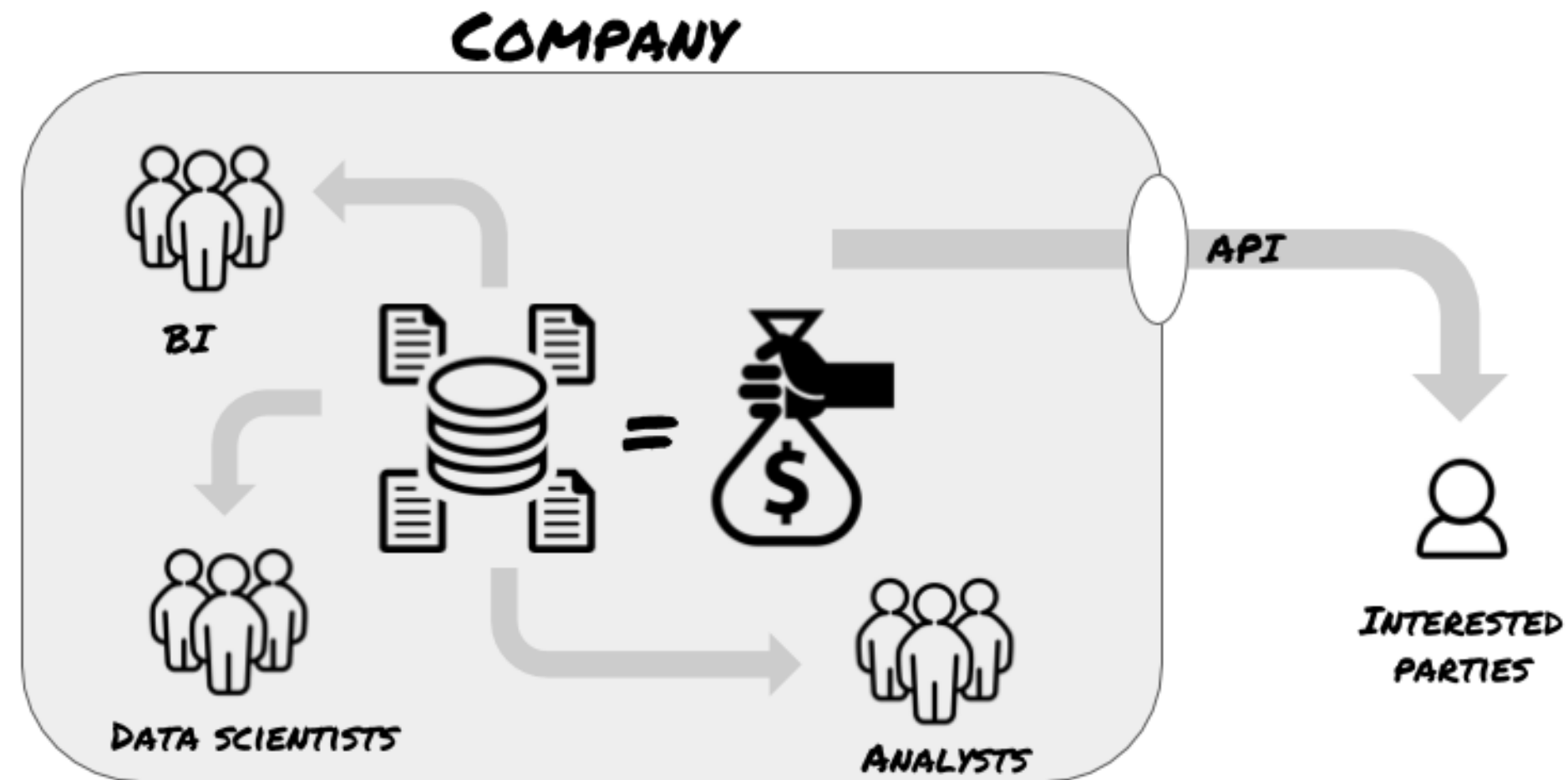
Course contents

- ingest data using Singer
 - apply common data cleaning operations
 - gain insights by combining data with PySpark
 - test your code automatically
 - deploy Spark transformation pipelines
- => intro to data engineering pipelines**

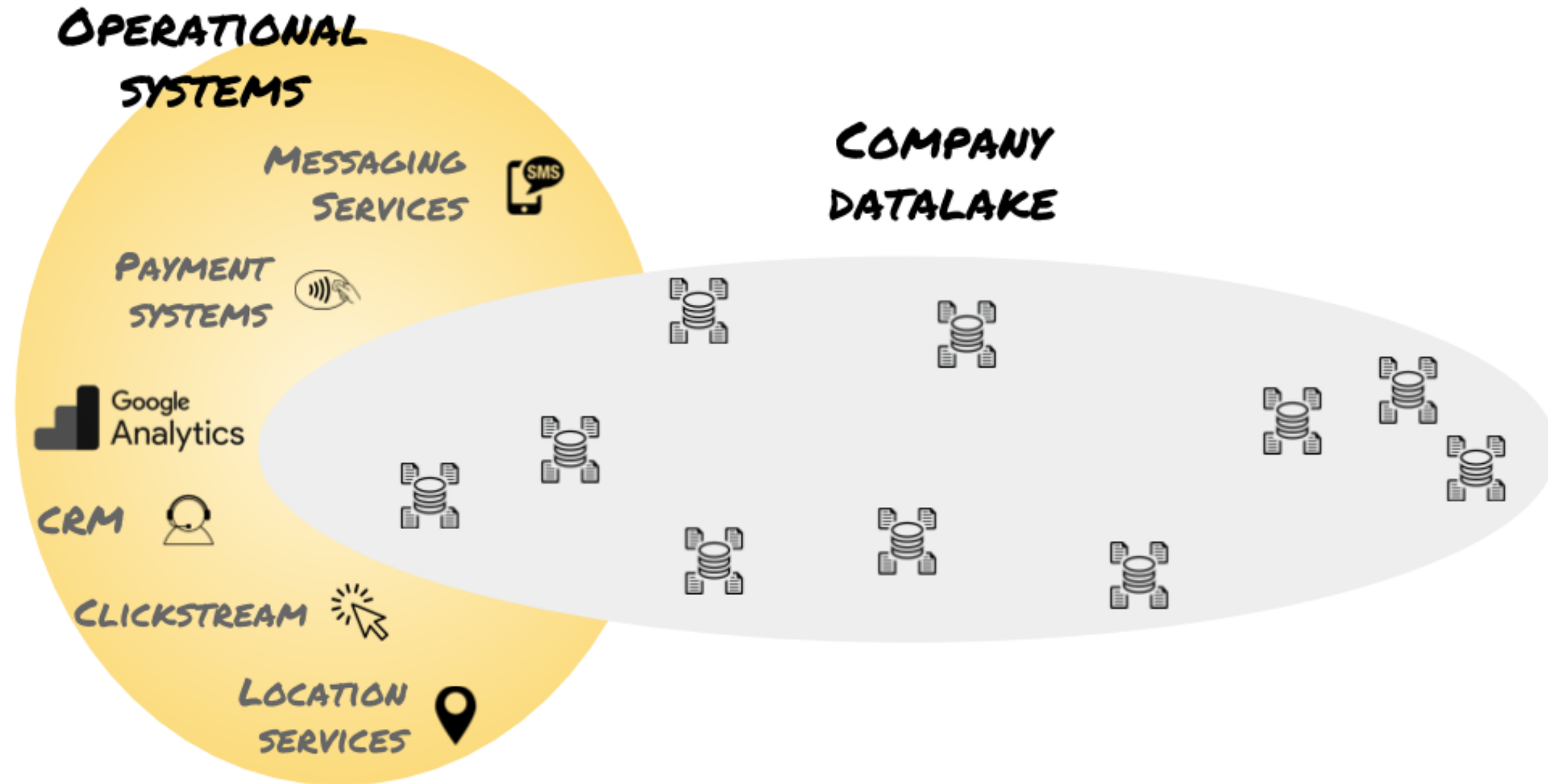
Data is valuable



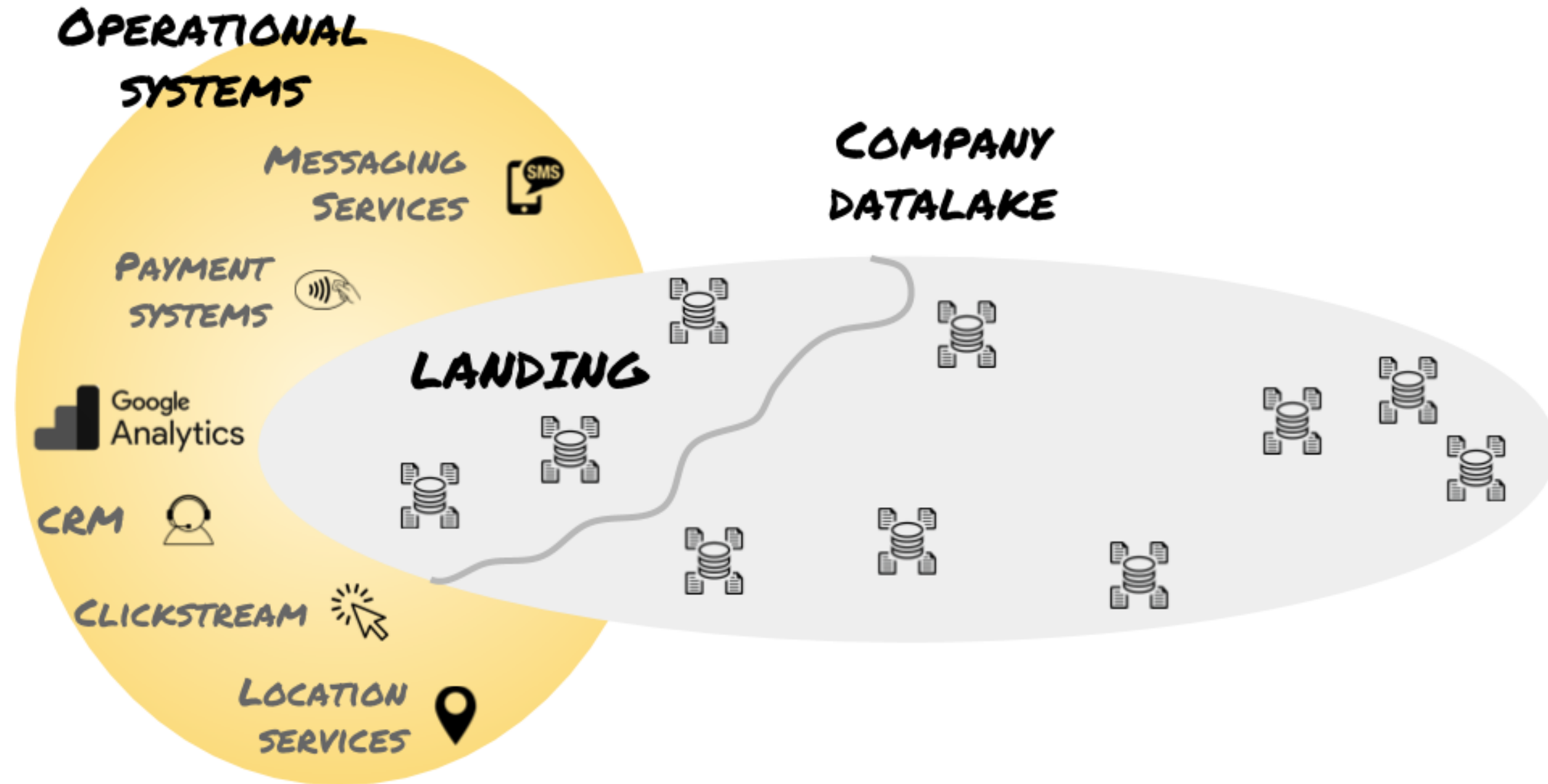
Democratizing data increases insights



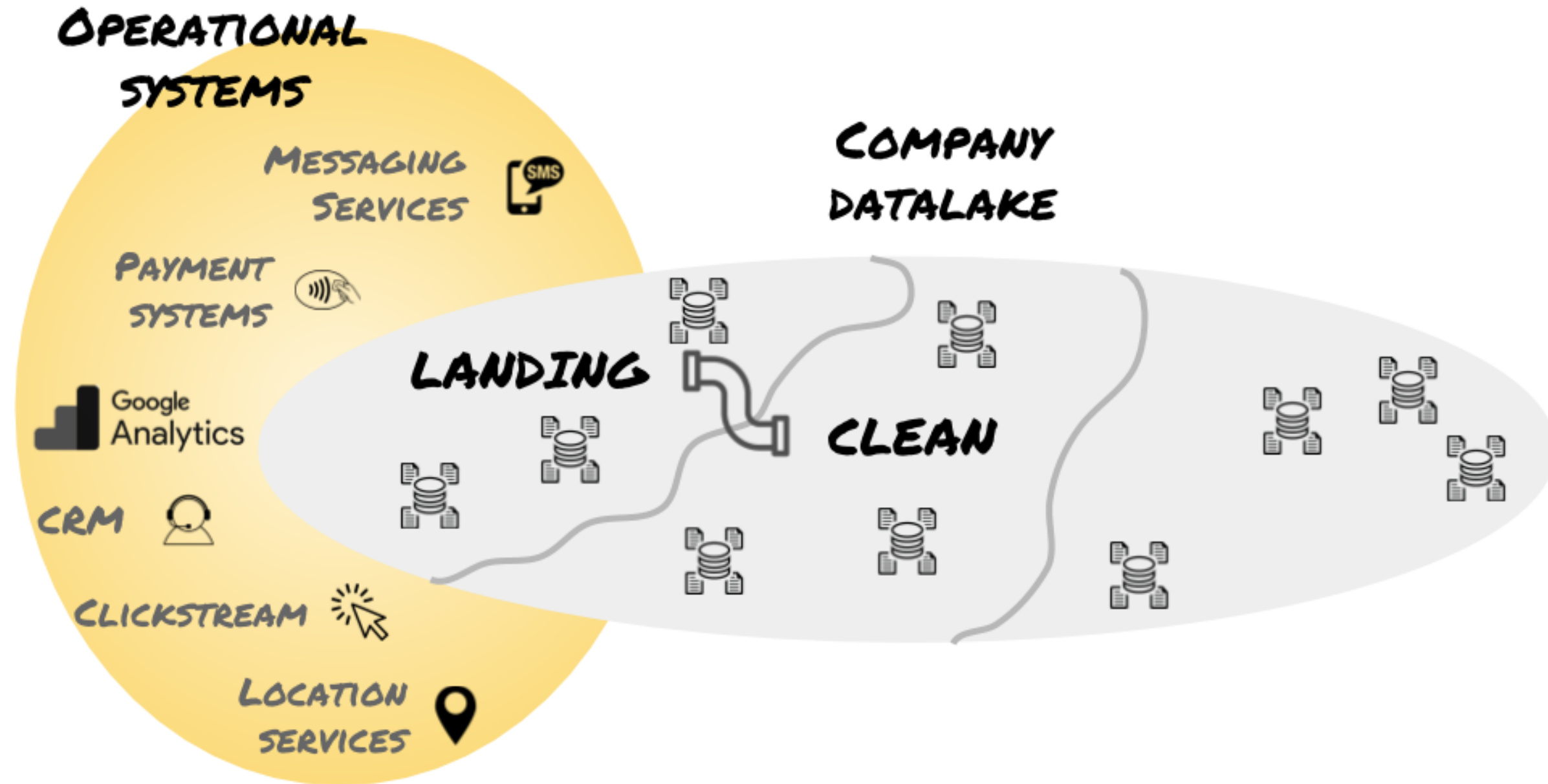
Genesis of the data



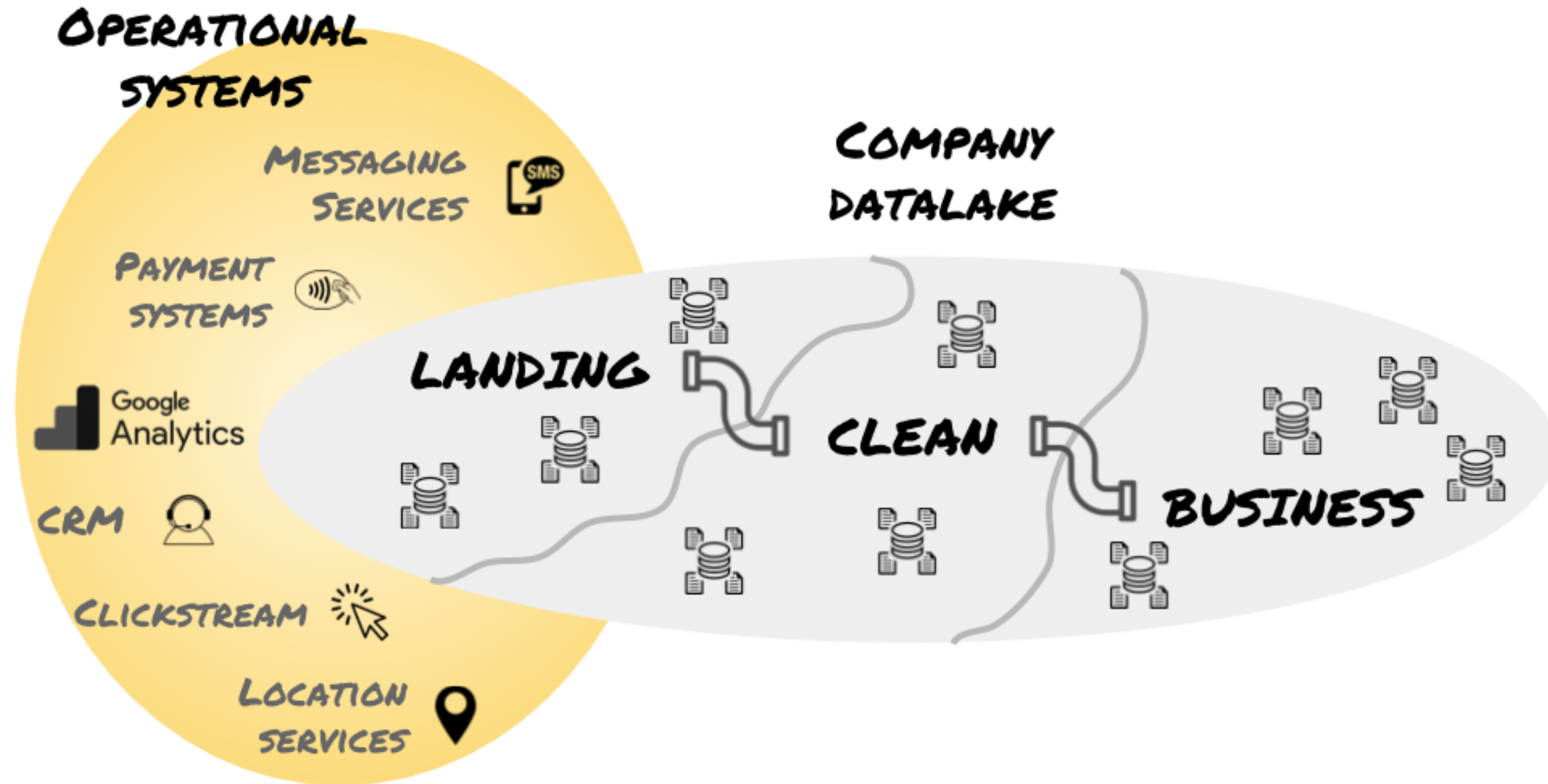
Operational data is stored in the landing zone



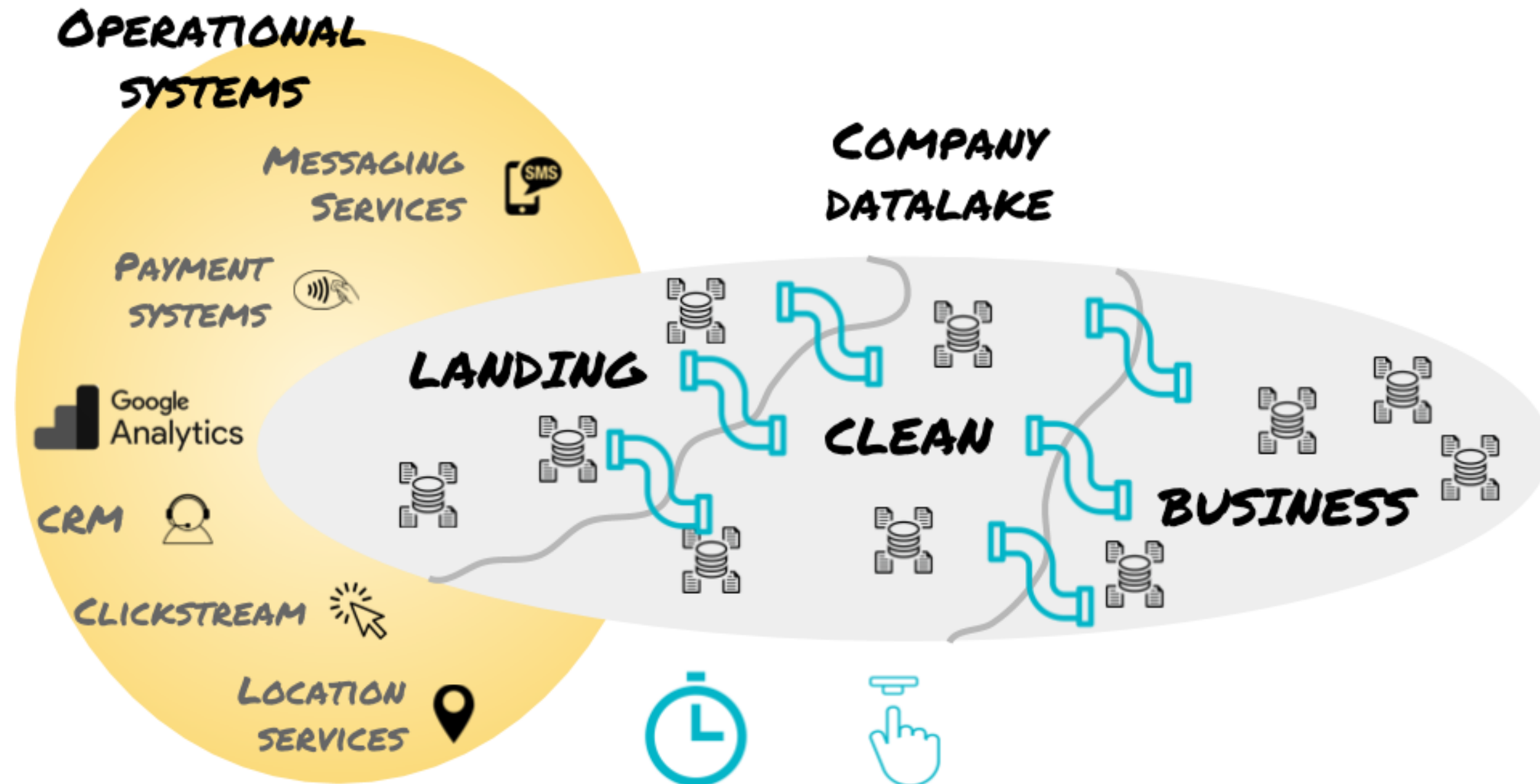
Cleaned data prevents rework



The business layer provides most insights



Pipelines move data from one zone to another



Let's reason!

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Introduction to data ingestion with Singer

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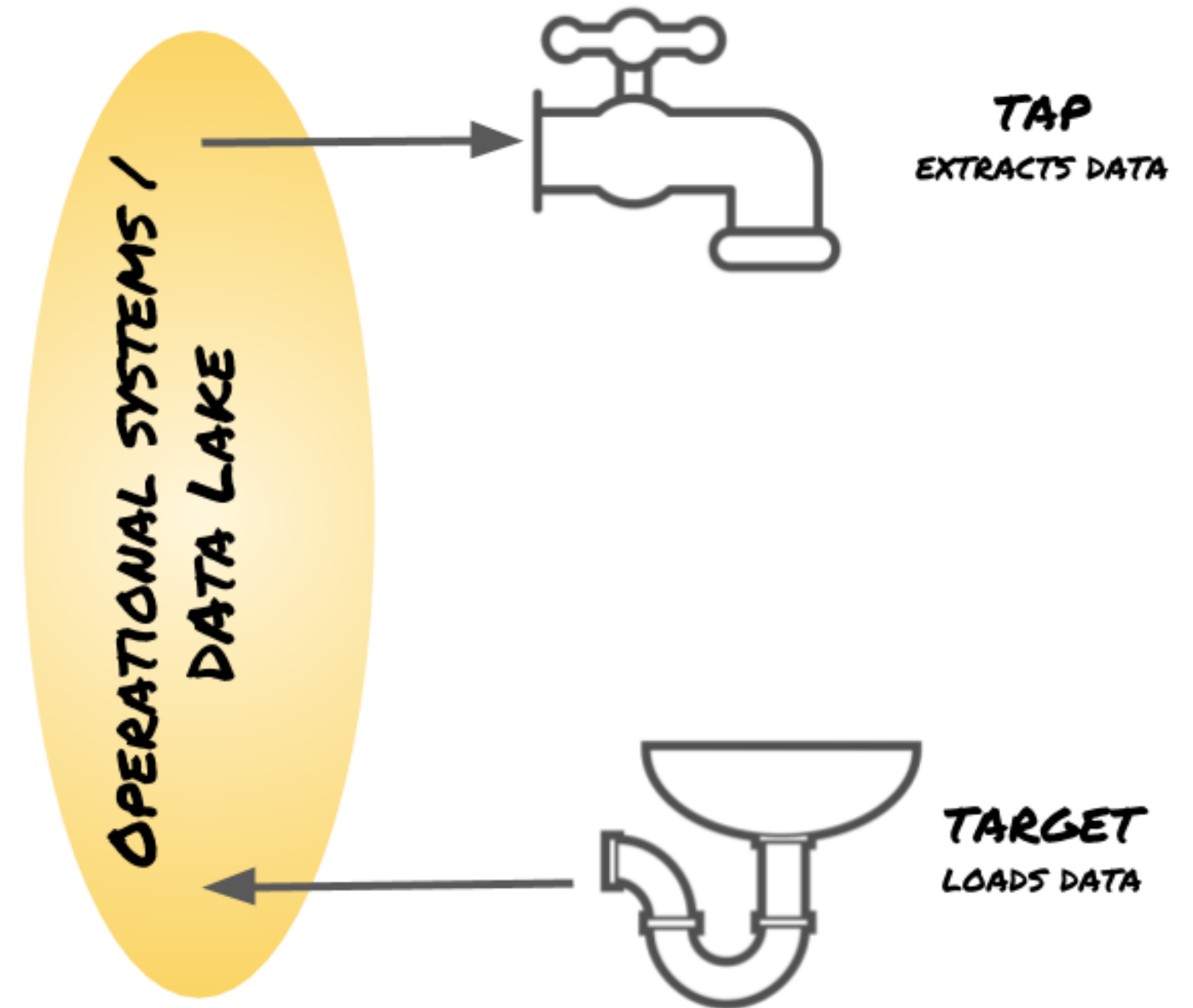
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Singer's core concepts

Aim: “The open-source standard for writing scripts that move data”

Singer is a *specification*

- data exchange format: *JSON*
- extract and load with *taps* and *targets*
 - => language independent

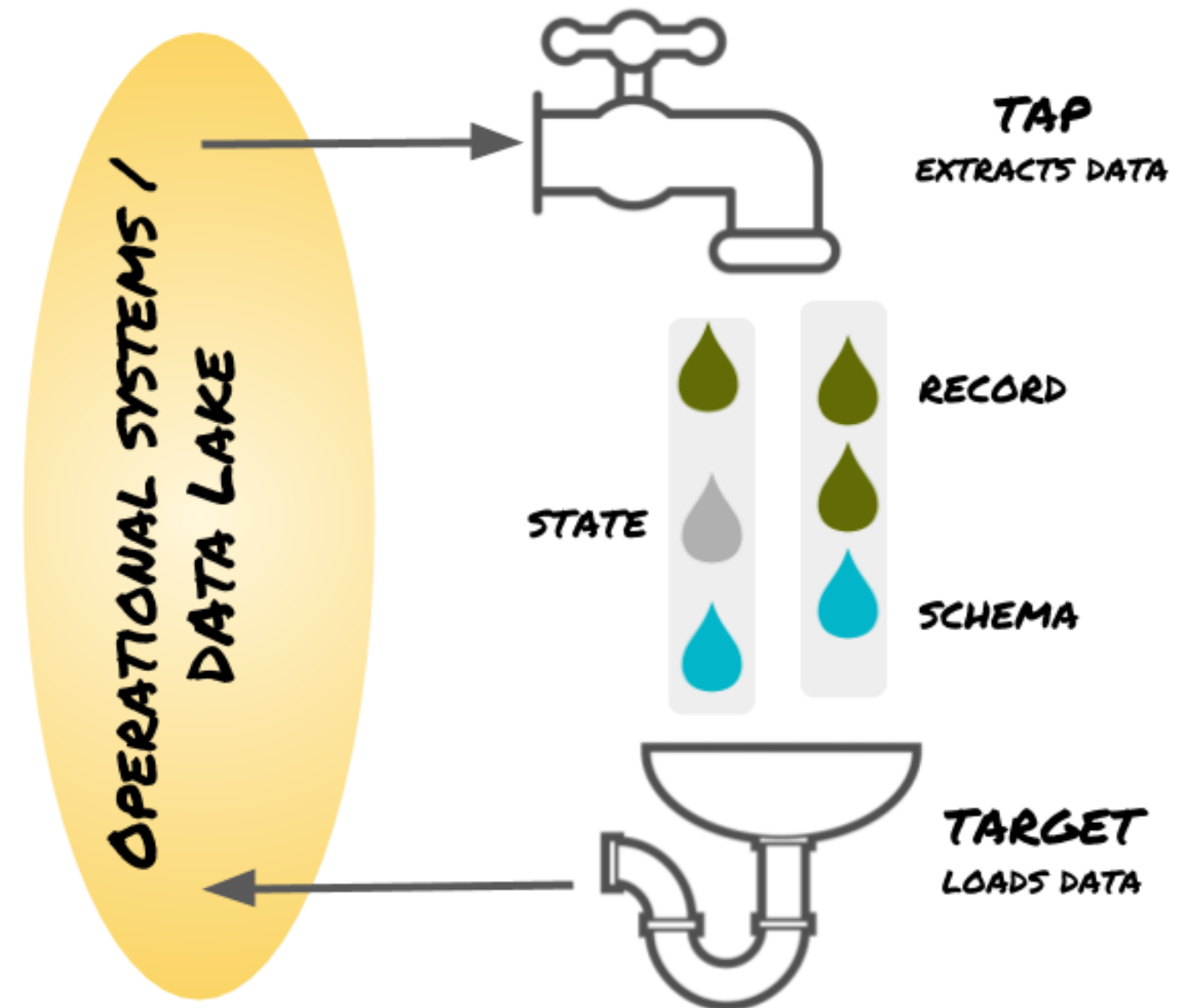


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- communicate over *streams*:
 - schema (metadata)
 - state (process metadata)
 - record (data)

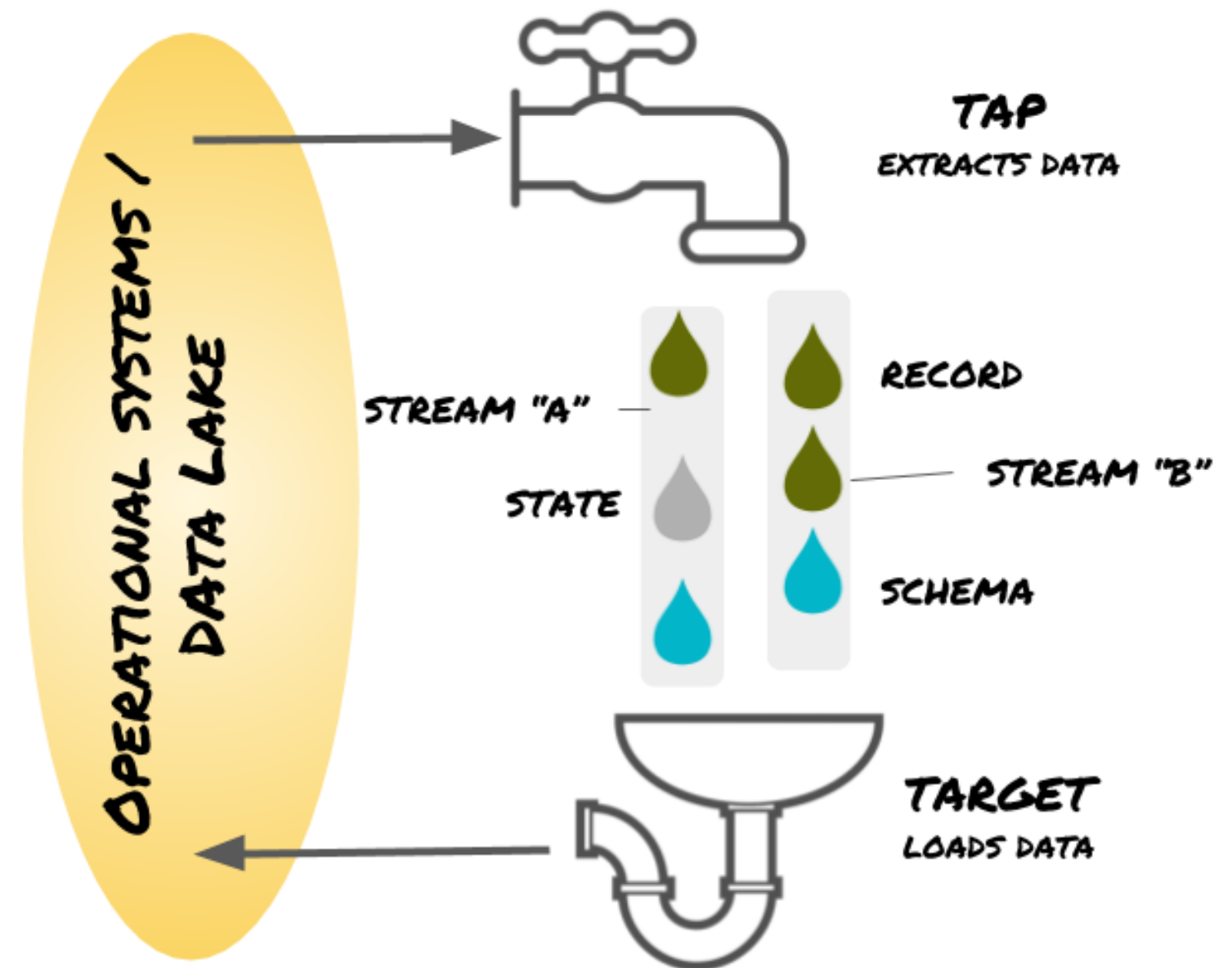


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Describing the data through its schema

```
columns = ("id", "name", "age", "has_children")
users = {(1, "Adrian", 32, False),
         (2, "Ruanne", 28, False),
         (3, "Hillary", 29, True)}

json_schema = {
    "properties": {"age": {"maximum": 130,
                           "minimum": 1,
                           "type": "integer"},
                  "has_children": {"type": "boolean"},
                  "id": {"type": "integer"},
                  "name": {"type": "string"}},
    "$id": "http://yourdomain.com/schemas/my_user_schema.json",
    "$schema": "http://json-schema.org/draft-07/schema#" }
```

Describing the data through its schema

```
import singer

singer.write_schema(schema=json_schema,
                    stream_name='DC_employees',
                    key_properties=["id"])
```

```
{"type": "SCHEMA", "stream": "DC_employees", "schema": {"properties":
{"age": {"maximum": 130, "minimum": 1, "type": "integer"}, "has_children":
{"type": "boolean"}, "id": {"type": "integer"}, "name": {"type": "string"}},
"$id": "http://yourdomain.com/schemas/my_user_schema.json",
"$schema": "http://json-schema.org/draft-07/schema#"}, "key_properties": ["id"]}
```


Serializing JSON

```
import json
```

```
json.dumps(json_schema["properties"]["age"])
```

```
'{"maximum": 130, "minimum": 1, "type": "integer"}'
```

```
with open("foo.json", mode="w") as fh:
```

```
    json.dump(obj=json_schema, fp=fh) # writes the json-serialized object  
                                     # to the open file handle
```

Let's practice!

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Running an ingestion pipeline with Singer

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Streaming record messages

```
columns = ("id", "name", "age", "has_children")
users = {(1, "Adrian", 32, False),
         (2, "Ruanne", 28, False),
         (3, "Hillary", 29, True)}
```

```
singer.write_record(stream_name="DC_employees",
                    record=dict(zip(columns, users.pop())))
```

```
{"type": "RECORD", "stream": "DC_employees", "record": {"id": 1, "name": "Adrian", "age": 32, "has_children": false}}
```

```
fixed_dict = {"type": "RECORD", "stream": "DC_employees"}
record_msg = {**fixed_dict, "record": dict(zip(columns, users.pop()))}
print(json.dumps(record_msg))
```

Chaining taps and targets

```
# Module: my_tap.py
import singer

singer.write_schema(stream_name="foo", schema=...)
singer.write_records(stream_name="foo", records=...)
```

Ingestion pipeline: **Pipe** the tap's output into a Singer target, using the `|` symbol (Linux & MacOS)

```
python my_tap.py | target-csv
python my_tap.py | target-csv --config userconfig.cfg
my-packaged-tap | target-csv --config userconfig.cfg
```

Modular ingestion pipelines

```
my-packaged-tap | target-csv
```

```
my-packaged-tap | target-google-sheets
```

```
my-packaged-tap | target-postgresql --config conf.json
```

```
tap-custom-google-scraper | target-postgresql --config headlines.json
```

Keeping track with state messages

Keeping track with state messages

id	name	last_updated_on
1	Adrian	2019-06-14T14:00:04.000+02:00
2	Ruane	2019-06-16T18:33:21.000+02:00
3	Hillary	2019-06-14T10:05:12.000+02:00

```
singer.write_state(value={"max-last-updated-on": some_variable})
```

Run this `tap-mydelta` on 2019-06-14 at 12:00:00.000+02:00 (2nd row wasn't yet present then):

```
{"type": "STATE", "value": {"max-last-updated-on": "2019-06-14T10:05:12.000+02:00"}}
```


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