datajudge

Express and test specifications against data from database

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Agenda

- 1. Introduction
- 2. Case study: using datajudge to validate Twitch data
 - i. Setup
 - ii. Expressing expectations
 - iii. Testing expectations
- 3. General scenarios for comparing data sources











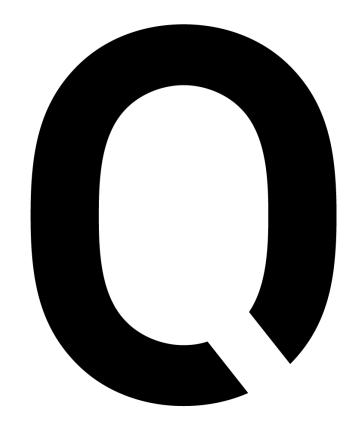






Me

- Computer Science at ETH Zurich and University of Washington
- Broad interest in Discrete Maths, Machine Learning and Software Engineering
- Currently Data Scientist/Machine Learning Engineer at QuantCo



quantco

- Data Science solutions
- Mostly fraud detection, pricing and demand forecasting
- join@quantco.com
- We contribute to OSS: conda,
 conda-forge, mamba, pandas,
 glum, ONNX, jupyter-lab

The data

channel	watch time	stream time	peak viewers	average viewers	followers	followers gained	
xQcOW	6196161750	215250	222720	27716	3246298	1734810	9
summit1g	6091677300	211845	310998	25610	5310163	1370184	8
Gaules	5644590915	515280	387315	10976	1767635	1023779	10
ESL_CSGO	3970318140	517740	300575	7714	3944850	703986	100
Tfue	3671000070	123660	285644	29602	8938903	2068424	7:
•••	•••	•••	•••	•••	•••	•••	

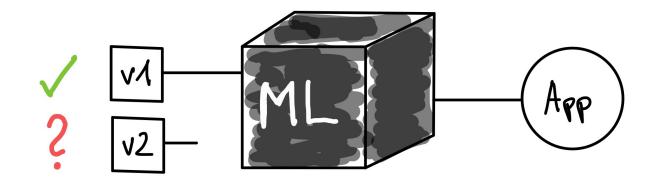
^{~1000} rows, 11 columns (7 numeric, 4 categorical)

Assumptions

- We rely on this data.
- We receive monthly data deliveries.
- Last month's delivery has been vetted manually.

The task

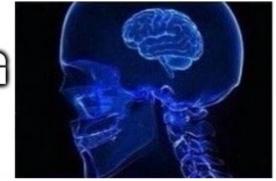
 We just received a new data delivery and want to use it for production purposes.

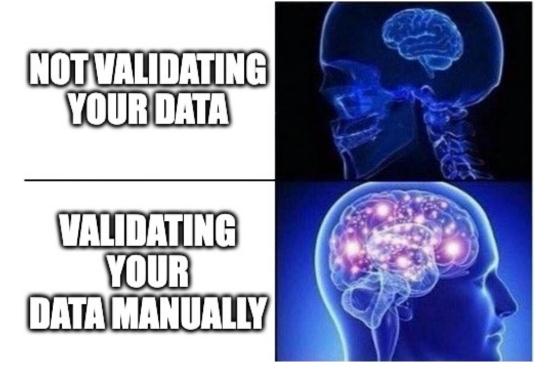


What could have happened between v1 and v2?

- Benign growth
 - 'Just more data', same distribution, no surprises, no expectations violated
- Malignant growth
 - Data drift
 - Bug in data-generating code
 - Failure to communicate change in requirements/business logic

NOTVALIDATING YOURDATA





DT VALIDATING YOUR DATA AUTOMATICALLY WITH Domain Knowledge

YOUR DATA AUTOMATICALLY WITH DOMAIN KNOWLEDGE AUTOMATICALLY WITH REFERENCE DATA

Using our domain knowledge to validate new data

• The language column should only contain values containing a non-empty sequence of the latin alphabet.

```
from datajudge import WithinRequirement
# Defining a data source.
within_requirement = WithinRequirement.from_table(
    table_name="twitch_v2",
    schema_name=schema_name,
    db name=db name,
# Defining a constraint on the data source.
within_requirement.add_varchar_regex_constraint(
    column="language",
    regex="^[a-zA-Z]+$",
```

Using old data to validate the new data

• The language column should contain the same unique values we already encountered before.

```
from datajudge import BetweenRequirement
between_requirement_version = BetweenRequirement.from_tables(
    db name1=db name,
    db_name2=db_name,
    schema_name1=schema_name,
    schema_name2=schema_name,
    table name1="twitch v1",
    table_name2="twitch_v2",
between_requirement_version.add_uniques_equality_constraint(
    columns1=["language"],
    columns2=["language"],
```

Using old data to validate new data

The column structure should be equal for both versions.

```
between_requirement_version.add_column_subset_constraint()
between_requirement_version.add_column_superset_constraint()
```

Using old data to validate new data

• The static features of the rows in the old data should be present and unchanged in the new data.

```
columns = ["channel", "partnered", "mature"]
between_requirement_version.add_row_subset_constraint(
    columns1=columns,
    columns2=columns,
    constant_max_missing_fraction=.01,
)
```

• Note: constant_max_missing_fraction is a tolerance parameter.

Using new data to validate new data

- The distribution of average_viewers should follow the same underlying data generating process in both versions.
- 2-sample Kolmogorov Smirnov hypothesis test

```
between_requirement_version.add_ks_2sample_constraint(
    column1="average_viewers",
    column2="average_viewers",
    significance_level=0.05,
)
```

Using new data to validate new data

• Mean of column average_viewers of mature channels shouldn't deviate too much from overall mean of that same column.

```
between_requirement_columns = BetweenRequirement.from_tables(
    db_name1=db_name,
    db_name2=db_name,
    schema_name1=schema_name,
    schema_name2=schema_name,
    table_name1="twitch_v2",
    table_name2="twitch_v2",
between_requirement_columns.add_numeric_mean_constraint(
    column1="average_viewers",
    column2="average_viewers",
    condition1=None,
    condition2=Condition(raw_string="mature IS TRUE"),
    max_absolute_deviation=0.1,
```

Execution of tests

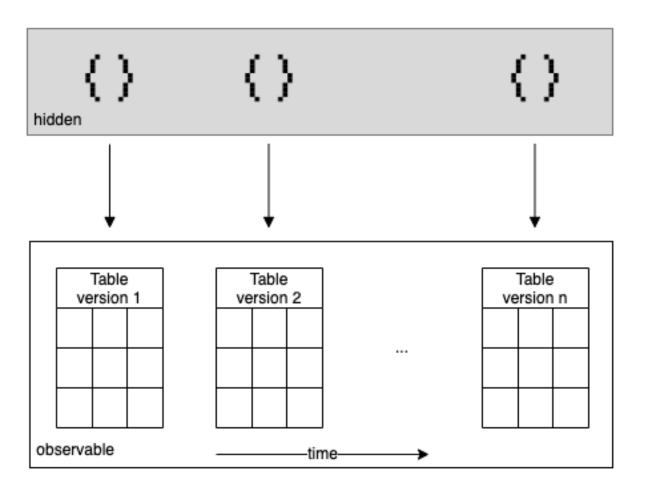
- All of the previous code simply goes into a Python file.
- Add few lines of boilerplate code:

```
requirements = [
    within_requirement,
    between_requirement_version,
    between_requirement_columns,
]
test_func = `**collect_data_tests(requirements)**`
```

• Execute with pytest: \$ pytest specification.py

AssertionError

```
test func[UniquesEquality::public.twitch v1 | public.twitch v2]
constraint = <datajudge.constraints.uniques.UniquesEquality object at 0x108087e20>
datajudge engine = Engine(postgresgl://datajudge:***@localhost:5432/datajudge)
@pytest.mark.parametrize(
"constraint", all_constraints, ids=Constraint.get_description
def test constraint(constraint, datajudge engine):
test result = constraint.test(datajudge engine)
        assert test result.outcome, test result.failure message
       AssertionError: tempdb.public.twitch_v1's column(s) 'language' doesn't have
        the element(s) '{'Sw3d1zh'}' when compared with the reference values.
/usr/local/Caskroom/.../lib/python3.10/site-packages/datajudge/pytest integration.py:25:
AssertionError
```

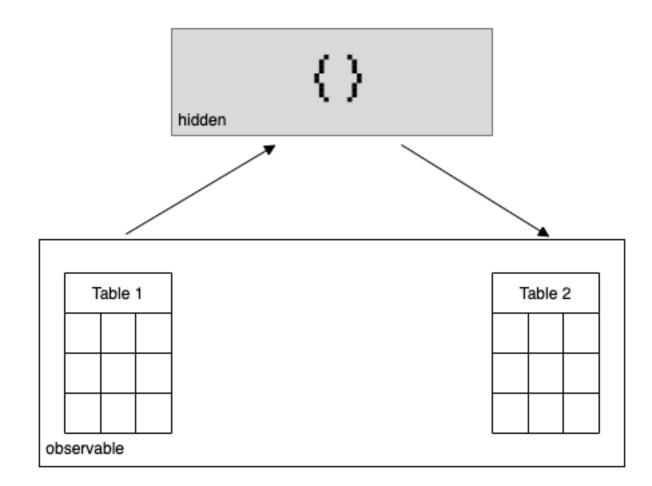


Comparisons between data sources: Across time

- Tables with equal structure and semantics.
- Tables from different points in time.
- Compare content with respect to expected change.
- Relevant use case: Obfuscated data-generating code between pipeline runs.

Comparisons between data sources: Across space

- Tables of different structure.
- Tables could have an input-output relationship.
- Compare tables with respect to invariants.
 - E.g. Min in summary stats table should be equal to min in unit table.
- Relevant use case: Obfuscated data-generating code between steps of a pipeline.



datajudge: Good to know

- Only relational databases are supported.
 - Currently test against Postgres, Snowflake and Mssql.
- Heavy lifting happens in database, only test results are fetched to memory.
 - Can be very advantageous in terms of memory consumption and runtime.
- datajudge generates SQL queries from high-level API.
 - It relies on the SQLAlchemy Language Expression API to be dialect-agnostic.
 - Generated queries are logged and can be used for data debugging.
- Many more constraints exist.
 - E.g.: constraints for date columns, useful to validate historization of database

One more thing

- datajudge has been collaborative work and grew as a side project.
- Please share thoughts/ideas/suggestions!
- https://github.com/QuantCo/datajudge

Questions?

How we use datajudge at QuantCo

- Generate html test reports with a pytest plugin.
 - pytest specification.py --html=report.html
 - Eases collaboration and archiving.
- Parametrize data sources.
 - pytest specification.py --new_db new_db --old_db old_db
- Subselect tests
 - pytest specification.py -k varchar_constraint
- Integration in Cl.

Why not just use database constraints?

- Not available across dialects/dbms.
- No error tolerance.
- No conditioning.
- No/hardly any comparisons between data sources.
- Only operate on a row-level.
 - E.g. constraining the mean of a column is not possible.
- Different workflow: pre-transaction vs. post-transaction.

Why not use **Great Expectations**?

	datajudge	Great Expectations
Comparisons between data	First-class citizen	Tricky:/
Data sources	Relational databases	Databases, files, in-memory data
Adoption	Fast startup time	Complex ecosystem
Objective	Testing	Exploration, monitoring, testing
SQL logic	By framework	By user/by framework