

Systems Integrations



May 13, 2020

OpenDP: An Open-Source platform for Differential Privacy



Library desiderata \Rightarrow System Principles

- Extensibility
- Flexibility
- Verifiability
- Programmability
- Modularity
- Usability
- Efficiency
- Utility

Key Questions

- What types of stores need Commons components integrate with?
- What types of **identity**, **authentification** and **authorization** systems will Commons integrate with?
- What are the use cases utilitizing distributed data/distributed computation?
- Are there limitations that language choices for library impose on systems?
- What is the best API for library adoption? (language neutral? unopinionated?)

Example System Joint with Microsoft

- This system is an end-to-end platform to allow DP queries and statistical releases to senstive data.
- We provide several basic building blocks that can be used by people involved with sensitive data, with implementations based on vetted and mature differential privacy research.
- We provide a pluggable open source framework that researchers can use to try newer mechanisms and algorithms.

Example System Joint with Microsoft



Components

Current Component List:

- 1. Library of DP algorithms, components, utilities
- 2. Analysis Graph API
- 3. **SDK** support to execute against CSV, SQL Server, PostgreSQL, Dataverse, in Azure
- 4. **SQL** an ODBC-style API to produce differentially private reports from SQL queries (SQL-92), focused on common aggregates used in analytics.
- 5. **Static Validator** certify a proposed analysis is DP and compose privacy loss
- 6. **Stocastic Validator** test workflows numerically against synthetic data for privacy violations

Library Language Choice

Library implemented in Rust:

- Rust is a fast-growing, imperative functional language.
- high performance, reliable, good memory safety and type safety.
- structures, generics, higher-order functions, and values are immutable by default.
- It has a strong ownership model that gives memory safety.

Collectively, this makes Rust code easier to reason about concretely.

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Proposal: Rust should be the language for the robust, polished, production code that enters the maximally trusted OpenDP library. However, to encourage contributions from researchers and other developers, the library should be able to work with Python and R code components.

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- API to the library defines an analysis as a computational graph.
- Proposed analysis is validated for privacy preservation.
- If certified, sent to the runtime.
- Python and R bindings allow easy access to specify an analysis through the API.

```
In [2]: income edges = list(range(0, 100000, 10000))
education categories = ["1", "2", "3", "4", "5", "6", "7", "8", "9", "10", "11", "12", "13", "14", "15", "16"]
with system. Analysis() as analysis:
    data = op.Dataset(path = data path, column names = var names)
    income histogram = op.dp histogram(
            op.cast(data['income'], type='int', lower=0, upper=100),
            edges = income edges,
            upper = 1000,
            null value = 150,
            privacy usage = { 'epsilon': 0.5}
    income prep = op, histogram(op, cast(data['income'], type='int', lower=0, upper=100),
                                     edges=income edges, null value =-1)
    income histogram2 = op.laplace mechanism/income prep. privacy usage={"epsilon": 0.5, "delta": .000001})
    sex histogram = op.dp histogram(
            op.cast(data['sex'], type='bool', true label="0"),
            upper = 1000.
            privacy usage = {'epsilon': 0.5}
    sex prep = op.histogram(op.cast(data['sex'], type='bool', true label="0"), null value = True)
    sex histogram2 = op.laplace mechanism(sex prep. privacy usage={"epsilon": 0.5. "delta": .000001})
    education histogram = op.dp histogram(
            data['educ'].
            categories = education categories.
            null value = "-1".
            privacy usage = { 'epsilon': 0.5}
    education prep = op.histogram(data['educ'],
            categories = education categories, null value = "-1")
    education histogram2 = op.laplace mechanism(education prep. privacy usage={"epsilon": 0.5, "delta": .000001})
analysis.release()
print("Income histogram Geometric DP release: " + str(income histogram.value))
print("Income histogram Laplace DP release:
                                               + str(income histogram2.value))
print("Sex histogram Geometric DP release:
                                                * + str(sex histogram.value))
print("Sex histogram Laplace DP release:
                                               + str(sex histogram2.value))
print("Education histogram Geometric DP release:" + str(education histogram.value))
print("Education histogram Laplace DP release: " + str(education histogram2.value))
Income histogram Geometric DP release: [337 198 137 78 46 60 61 69 58 771
Income histogram Laplace DP release:
                                         1328.17803145 189.80390631 122.94969197 96.43640724 60.06534022
  AD EATCACEE AC ATLCACO 10 (1147614 16 (00000000 CT 075400005)
```



- We built a library API using Protocol Buffers,
 - serialized data language, like JSON or XML, but heavily typed, very compact.
- Rust uses same Foreign Function Interfaces as C++ so bindings direct to Rust are identical (cleaner to write),
- API provides bindings to Python and R,
- For Python, up on PyPI and pip installable.



The following contain examples of the computational graph representation of an analysis (or plan) for a proposed release.















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