Standardizing MPC for Privacy Preserving Measurement

RWC 2022

Tim Geoghegan Eric Rescorla Christopher Patton Christopher Wood



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- ENPA designed to solve a specific use case ⇒ What to do about aggregates for which ENPA/Prio is not well-suited?
- Cryptography is advancing (lots of MPC protocols in the literature for various aggregation functions) ⇒ Lacks a clear roadmap for deployment

• The IETF has formed a <u>working group</u> for standardizing *privacy-preserving measurement*.

charter-ietf-ppm-01

There are many situations in which it is desirable to take measurements of data which people consider sensitive. For instance, a browser company might want to measure web sites that do not render properly without learning which users visit those sites, or a public health authority might want to measure exposure to some disease without learning the identities of those exposed. In these cases, the entity taking the measurement is not interested in people's individual responses but rather in aggregated data (e.g., how many users had errors on site X). Conventional methods require collecting individual measurements in plaintext and then aggregating them, thus representing a threat to user privacy and rendering many such measurements difficult and impractical.

New cryptographic techniques address this gap through a variety of approaches, all of which aim to ensure that the server (or multiple, non-colluding servers) can compute the aggregated value without learning the value of individual measurements. The Privacy Preserving Measurement (PPM) work will standardize protocols for deployment of these techniques on the Internet. This will include mechanisms for:

- Client submission of individual measurements, potentially along with proofs of validity $% \left({{{\boldsymbol{x}}_{i}}} \right)$

- Verification of validity proofs by the server(s), if sent by client

- Computation of aggregate values by the server(s) and reporting of results to the entity taking the measurement

A successful PPM system assumes that clients and servers are configured with each other's identities and details of the types of measurements to be taken. This is assumed to happen out of band and will not be standardized in this WG.

The WG will deliver one or more protocols which can accommodate multiple PPM algorithms. The initial deliverables will support the calculation of simple predefined statistical aggregates such as averages, as well as calculations of the values that most frequently appear in individual measurements. The PPM protocols will use cryptographic algorithms and protocols defined by the CFRG to enable privacy-preserving properties. The protocol will be designed to limit abuse by both client and server, including exposure of individual user measurements and denial of service attacks on the measurement system. The resulting document(s) shall consider deployment contexts, and clearly describe abuse cases and remaining attacks which are not prevented or mitigated by the protocol(s).

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 - Out-of-scope: General-purpose MPC

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Overview

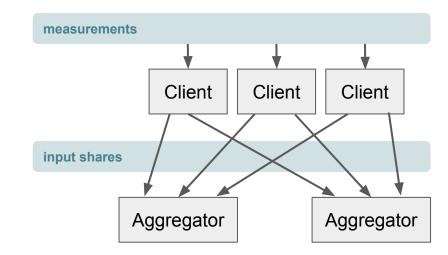
Part I

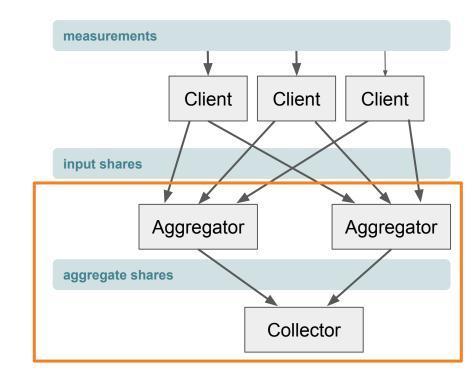
- Prio [CGB17]
- Poplar [BBCG+21]
- Other candidates

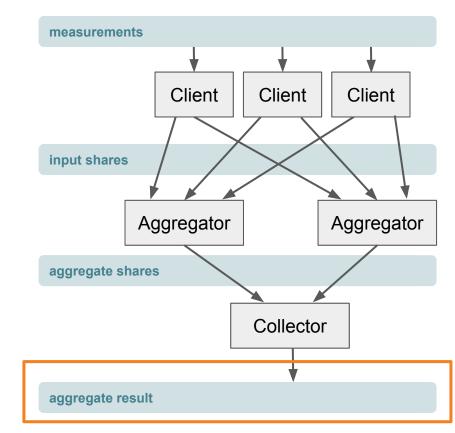
Part II

- Verifiable Distributed Aggregation Functions (VDAFs)
- The Privacy-Preserving Measurement (PPM) protocol

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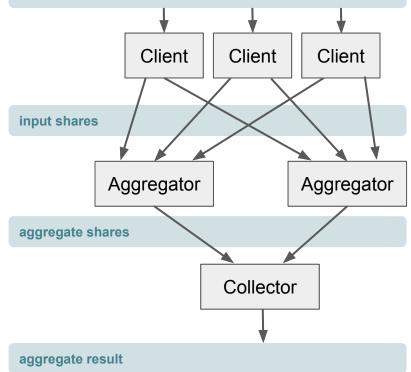




E.g.: Are users of my website experiencing high-latency?

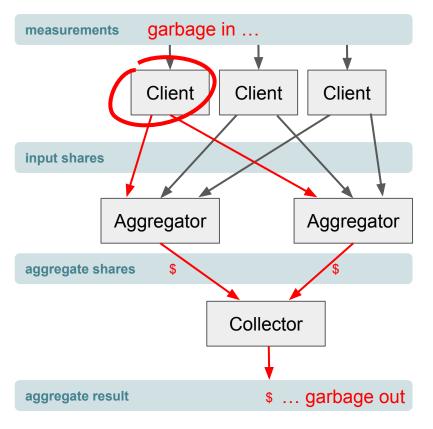
type	measurements	aggregate result
Count (<u>ENPA</u>) (high latency?)	1, 1, 0, 1, 0, 1	5
Mean, standard deviation (of load time)	182, 160, 190, 170, 175	175, 11
Histogram (estimating distribution of load time)	$\begin{array}{c} -7 \Rightarrow [1, 0, 0] \\ 23 \Rightarrow [0, 1, 0] \\ 45 \Rightarrow [0, 1, 0] \\ 59 \Rightarrow [0, 0, 1] \end{array}$	2.0 1.5 1.0 0.5 0.0 < 0 < 50 ≥ 50
Linear regression (load time as a function of no. of hops from client to server)	(1, 7), (2, 10), (3, 9), (4, 11), , (5, 10)	$\begin{array}{c} 12\\ 10\\ 8\\ 6\\ 4\\ 2\\ 0\\ 1\\ 2\\ 3\\ 4\\ 5 \end{array}$

measurements



Privacy Preserving Measurement (PPM) – RWC 2022

type	measurements	aggregate result
Count (<u>ENPA</u>) (high latency?)	1, 1, 0, 1, 0, <mark>999</mark>	1002
Mean, standard deviation (of load time)	182, 160, 190, 170, <mark>999</mark>	340, 368
Histogram (estimating distribution of load time)	$\begin{array}{l} -7 \Rightarrow [1, 0, 0] \\ 23 \Rightarrow [0, 1, 0] \\ 45 \Rightarrow [0, 1, 0] \\ [999, 999, 999] \end{array}$	1250 1000 750 500 250 0 < 0
Linear regression (load time as a function of no. of hops from client to server)	(1, 7), (2, 10), (3, 9), (4, 11), , <mark>(999, -999)</mark>	250 0 -250 -500 -1000 0 250 500 750

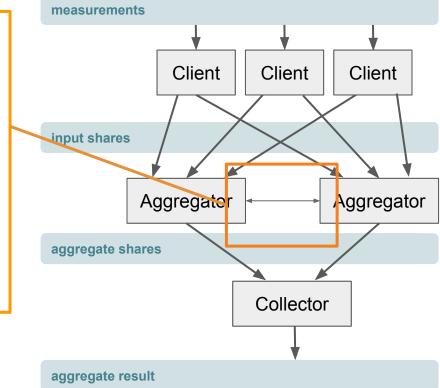


- Each measurement type specifies an **arithmetic circuit** C that recognizes valid inputs
- Each client generates a *fully linear proof (FLP)* [BBCG+19] of its input's validity
 - Proof shares allow Aggregators to jointly evaluate C on the secret shared input

 (X_1)

 (X_2)

1



- **Problem** securely aggregate the *heavy hitters*
 - Measurements: Arbitrary, bit strings
 - Aggregate result: Strings with at least T *hits*



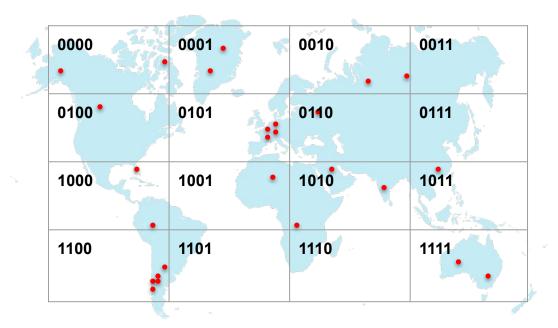
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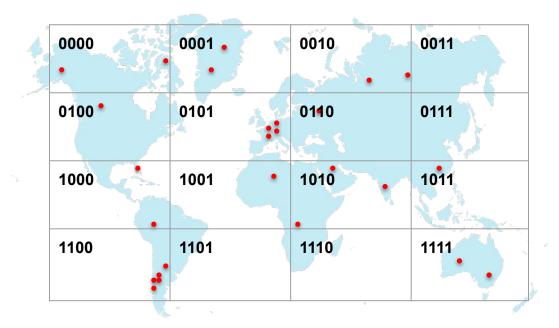
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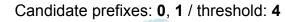
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 - Counts the number of strings beginning with a given *prefix*

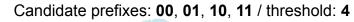


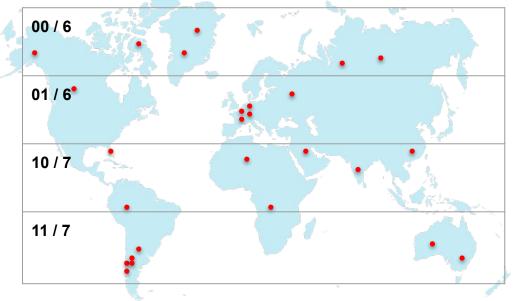
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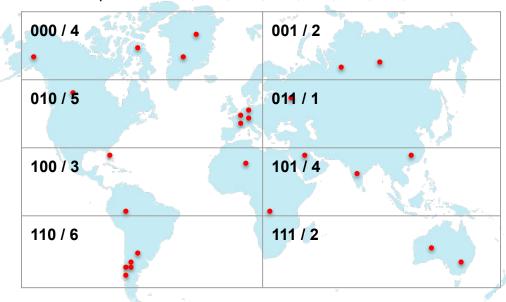
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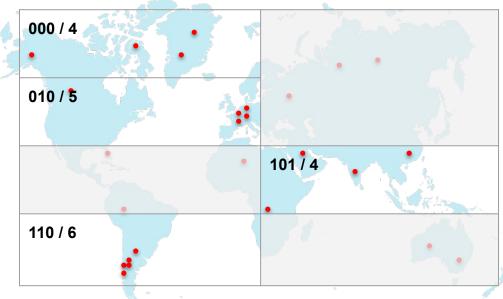


Candidate prefixes: 000, 001, 010, 011, 100, 101, 110, 111 / threshold: 4

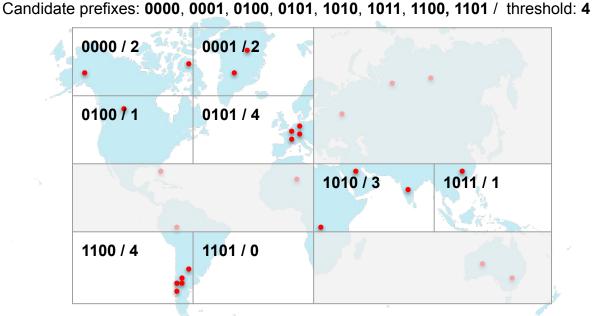
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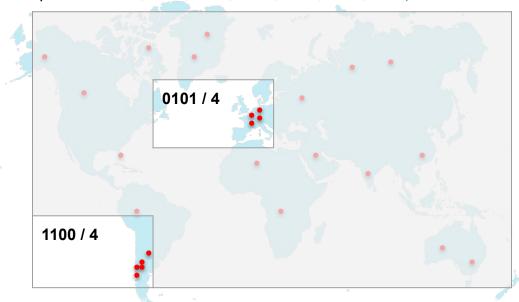
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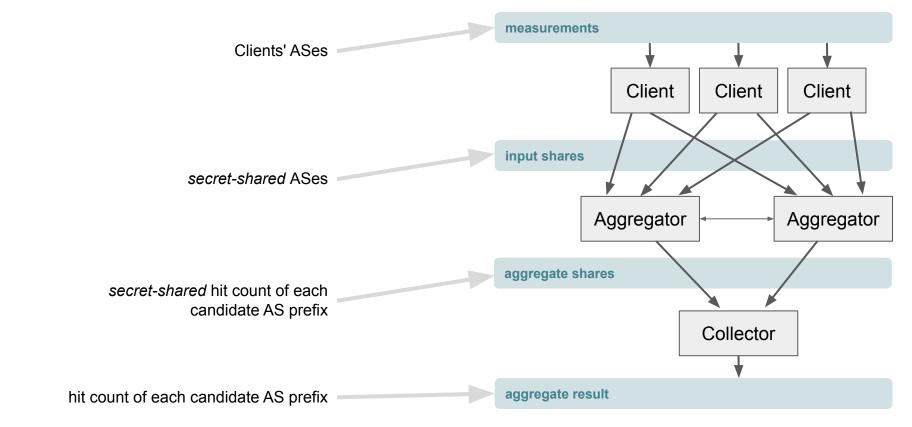


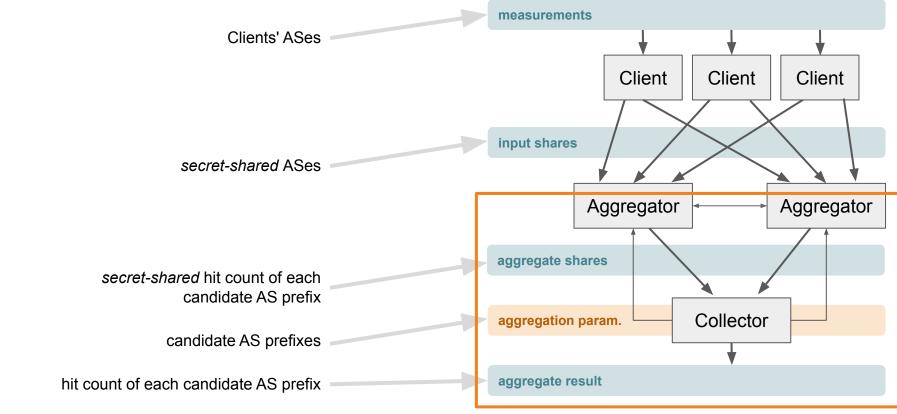
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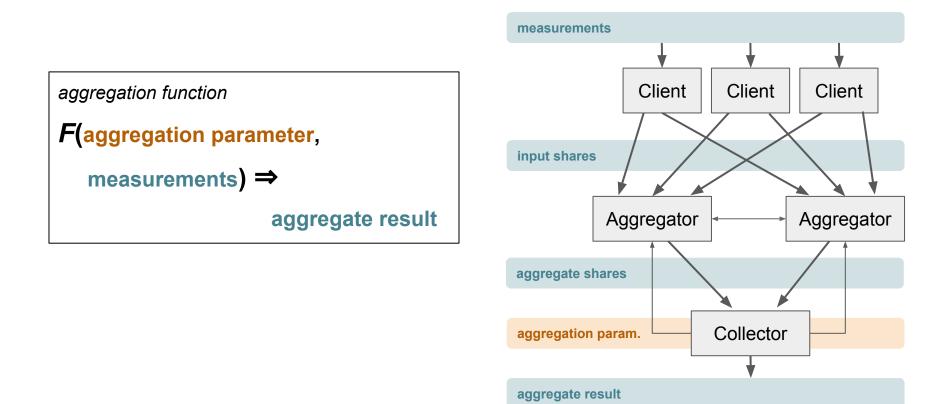






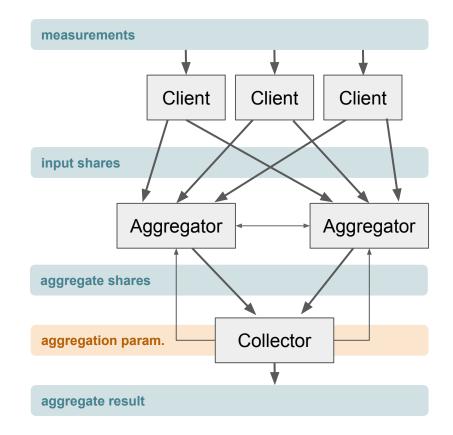
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Other candidates



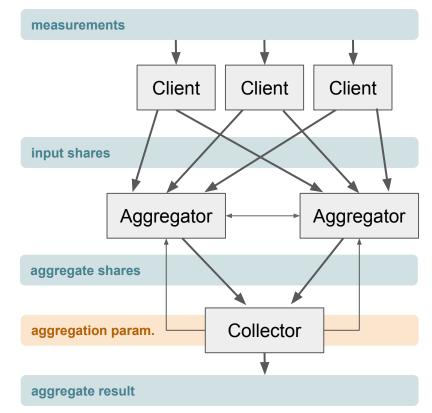
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 - Boolean-to-arithmetic conversion via OT extension
 - Compared to Prio, this significantly reduces Client computation for certain measurement types.



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 - Boolean-to-arithmetic conversion via OT extension
 - Compared to Prio, this significantly reduces Client computation for certain measurement types.
- Masked LARk [PCG+21]
 - Compute gradient descent over plaintext features and "masked" labels
 - Challenge Private, verifiable gradient descent computation that fits this architecture

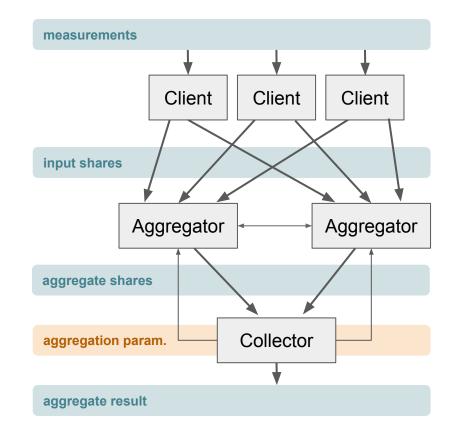


One scheme to rule them all? Nope.

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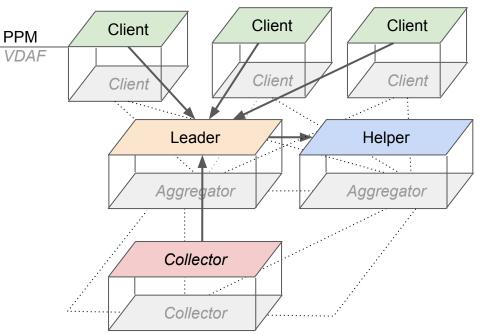
Verifiable Distributed Aggregation Functions (VDAFs)

- draft-patton-cfrg-vdaf-01
 - Defines syntax and (informal) security goals for VDAFs
 - Specifies two constructions: Prio and Poplar



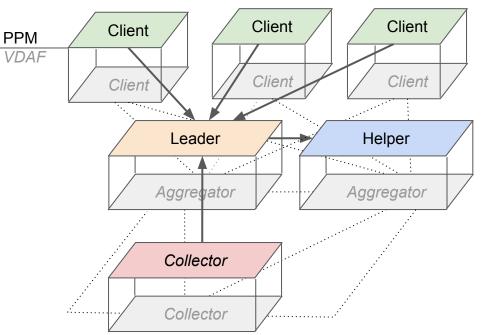
4. The Privacy-Preserving Measurement (PPM) protocol

• <u>draft-gpew-priv-ppm-01</u> – A protocol for evaluating a VDAF over HTTPS



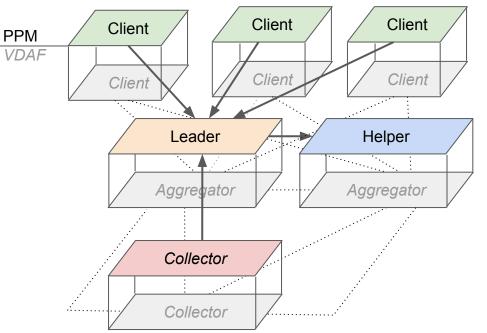
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 - Addresses a variety of operational issues (establishing secure channels, data recovery, picking a VDAF to run, etc.)
 - Additional security considerations:
 - Optional defenses against Sybil attacks
 - Support for differential privacy



How to contribute

- Join the PPM mailing list: ppm@ietf.org
- Provide feedback on:
 - draft-patton-cfrg-vdaf-01 (VDAF)
 - draft-gpew-priv-ppm-01 (PPM)
- Got an interesting paper, or a use case you're wondering about? Bring it to the list!

References

[AGJ+21] Addanki et al. "Prio+: Privacy Preserving Aggregate Statistics via Boolean Shares." ePrint #2021/576.

[CGB17] Corrigan-Gibbs-Boneh. "Prio: Private, Robust, and Scalable Computation of Aggregate Statistics." NSDI 2017.

[BBCG+19] Boneh et al. "Zero-Knowledge Proofs on Secret-Shared Data via Fully Linear PCPs." CRYPTO 2019.

[BBCG+21] Boneh et al. "Lightweight Techniques for Private Heavy Hitters". IEEE S&P 2021.

[PCG+21] Pfeiffer et al. "Masked LARk: Masked Learning, Aggregation and Reporting worKflow." arXiv:2110.14794.

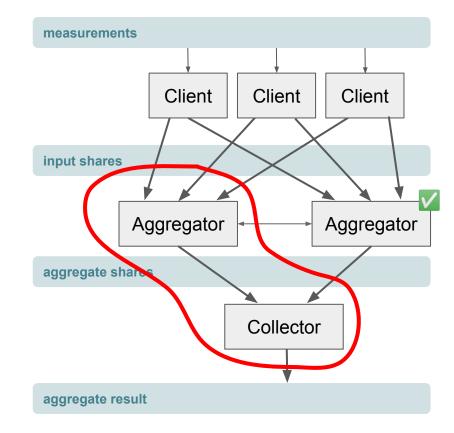


Backup slides

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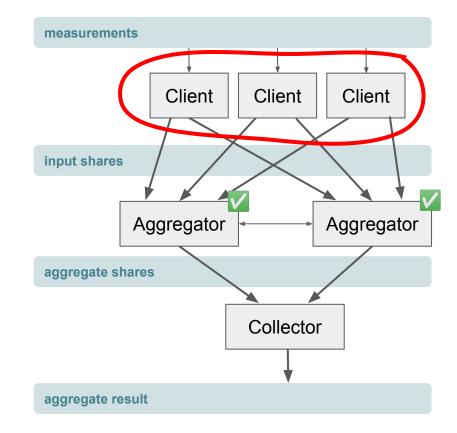
Security Requirements

• **Privacy** – If at least one Aggregator is honest, then no server learns anything beyond the aggregate result.



Security Requirements

- **Privacy** If at least one Aggregator is honest, then no server learns anything beyond the aggregate result.
- **Correctness** If all Aggregators implement the protocol correctly, then the Collector correctly computes the aggregate result over measurements uploaded by honest clients.



Contributors to VDAF/PPM drafts

Richard Barnes David Cook Armando Faz-Hernández Tim Geoghegan Charlie Harrison Anthony Miyaguchi Brandon Pitman Christopher Patton Eric Rescorla Phillipp Schoppmann Christopher Wood