

JOINT COMMUNICATION AND COMPUTING A PATH TOWARDS SUSTAINABLE IOT

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Assoc. Professor, Department of Electrical and Computer Engineering, AU

- MSc and PhD from DTU in Telecommunications
- Leading IoT WP in DIGIT research centre at AU
 - Sensor data compression, efficient transmission and storage, as well as approximate, real-time data analytics on compressed data
 - Mobile Edge Computing and Edge Intelligence for time-sensitive IoT applications
 - Lightweight sensor data encryption and privacy preserving schemes for IoT
- Currently leading three DFF projects “**AgileIoT**: Agile Edge Intelligence for Delay Sensitive IoT”, “**Light-IoT**: Analytics Straight on Compressed IoT Data”, “**eTouch**: Edge Intelligence for Immersive Telerobotics in Touch-enabled Tactile Internet”, and Horizon Europe MSCA-DN **TOAST** project.
- Participating in DFF **Growlean** project and H2020 ITN **IoTalentum** project.

OUTLINE

Motivation: why bridging communication with computing?

How to bridge communication with computing?

- Joint compression and encryption
 - Compressive sensing-based data encryption
- Joint compression and analytics
 - Analytics over GD compressed data

Conclusion and outlook

MOTIVATION

Enable a sustainable growth in IoT through a holistic design: Joint communication with computing



Wind turbine system



Autonomous vehicles



Industry 4.0



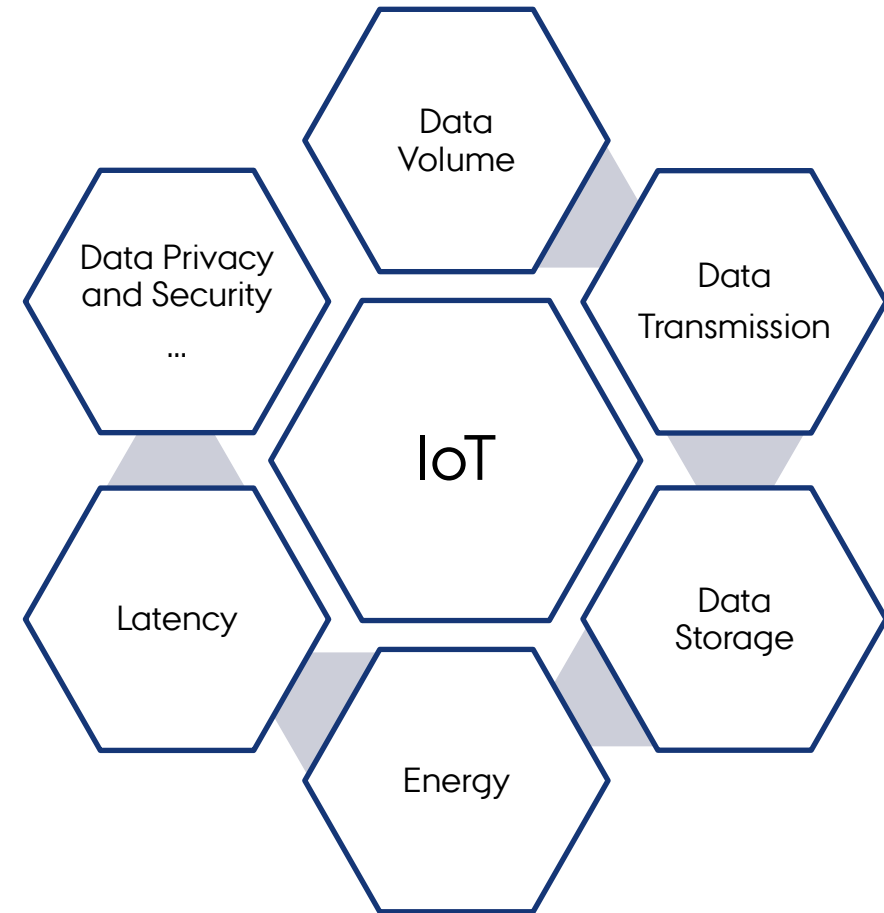
E-health



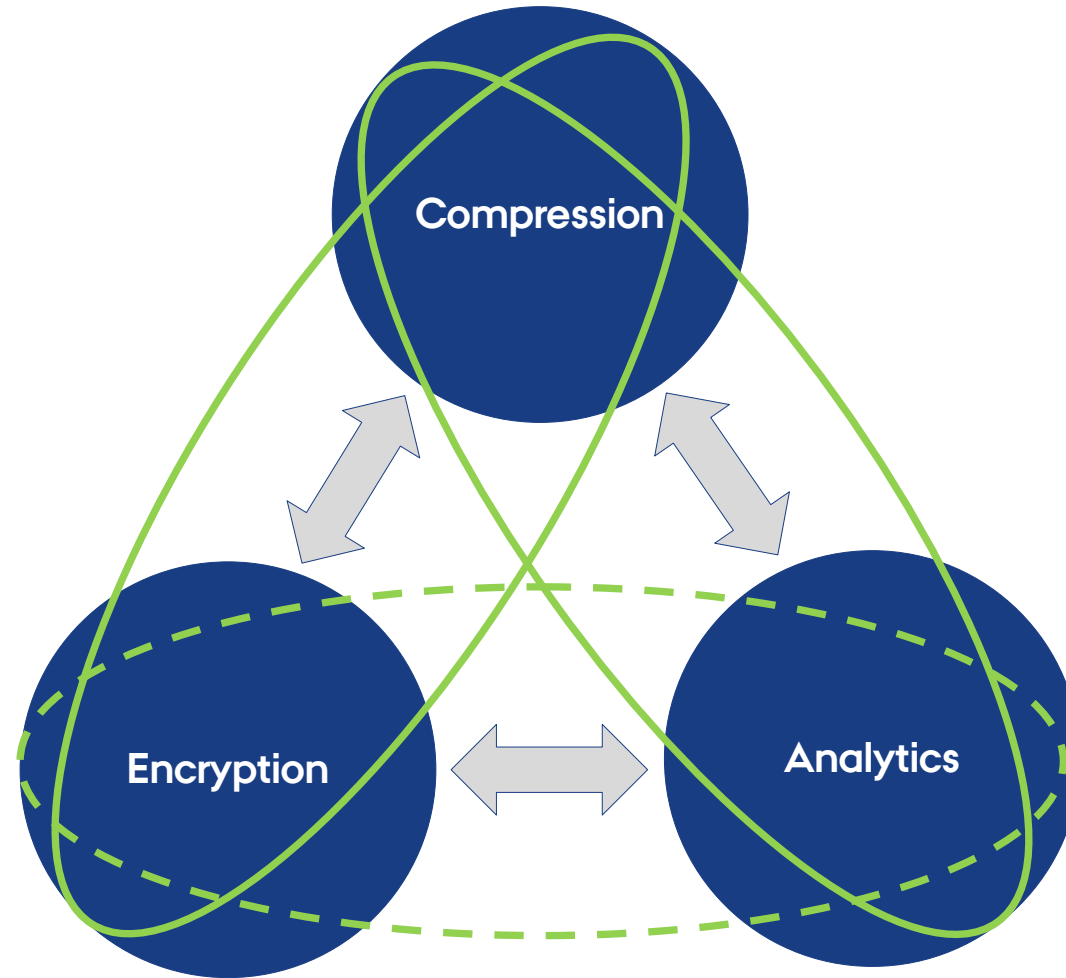
Smart agriculture



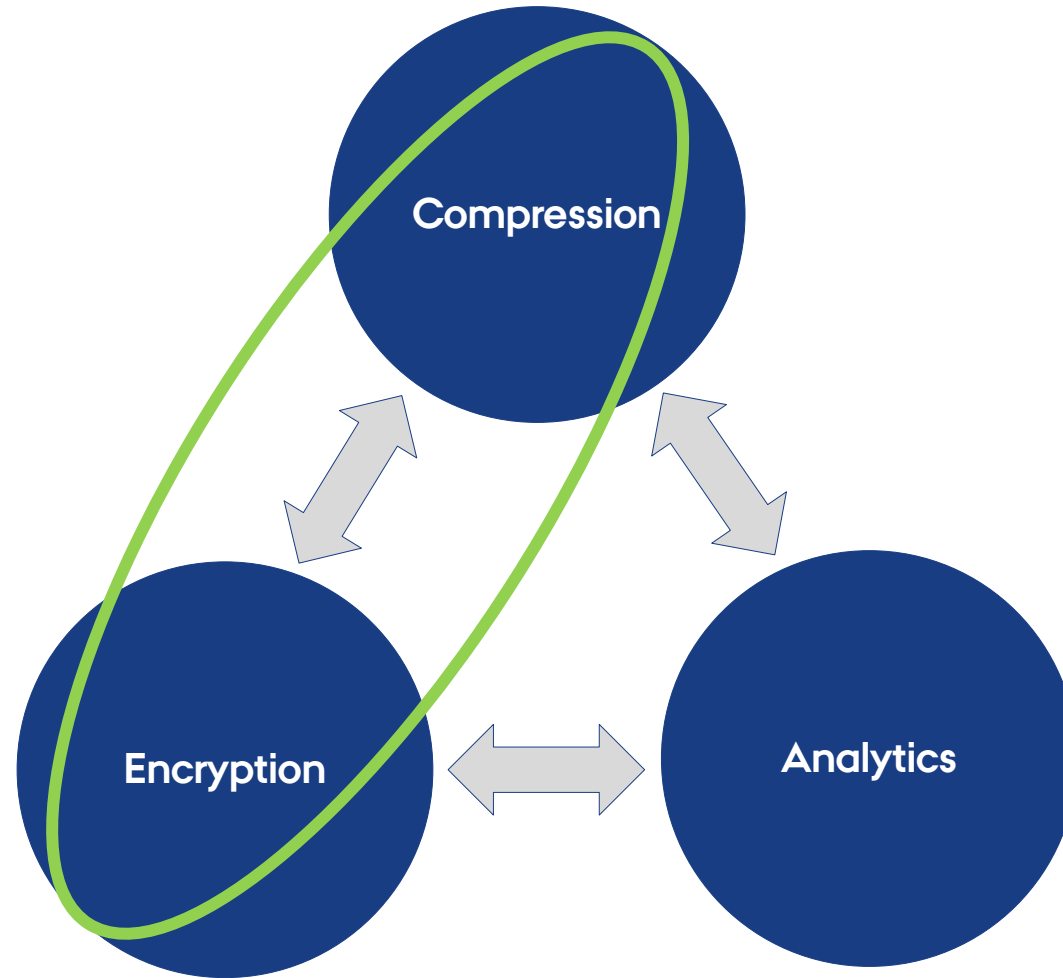
Smart city



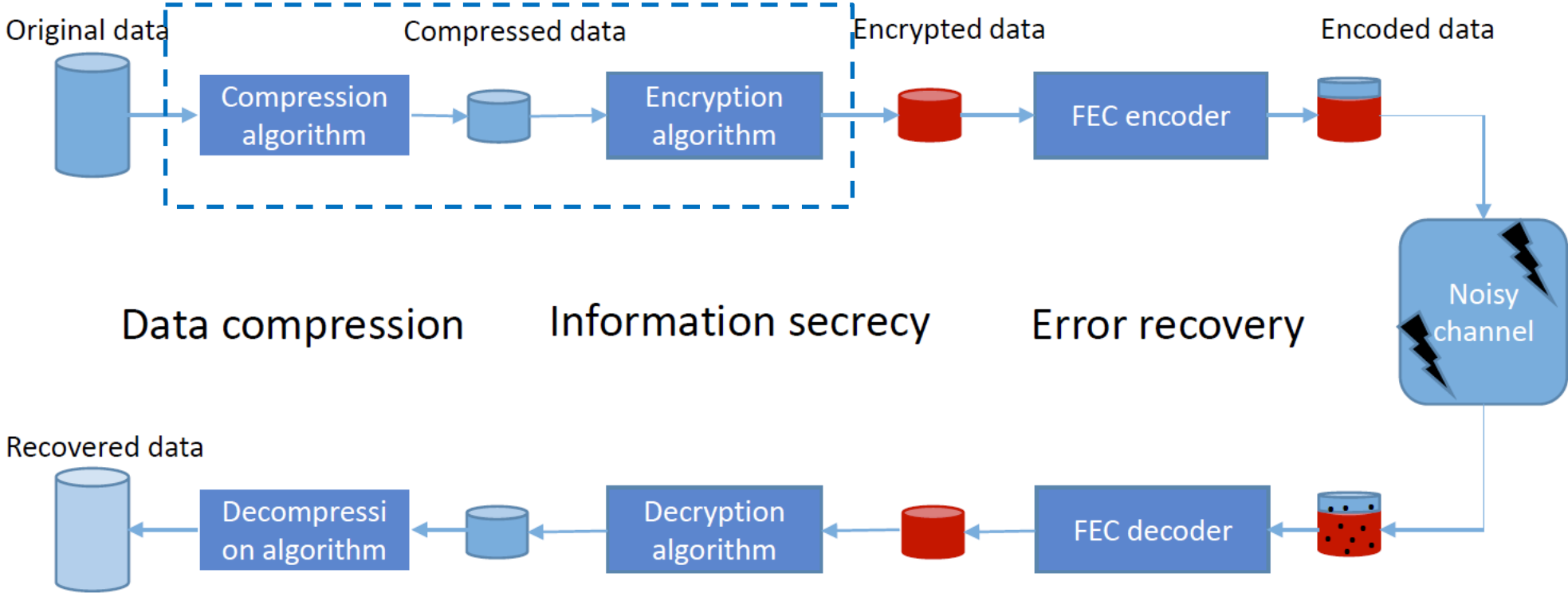
JOINT COMMUNICATION AND COMPUTING



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JOINT COMPRESSION AND ENCRYPTION

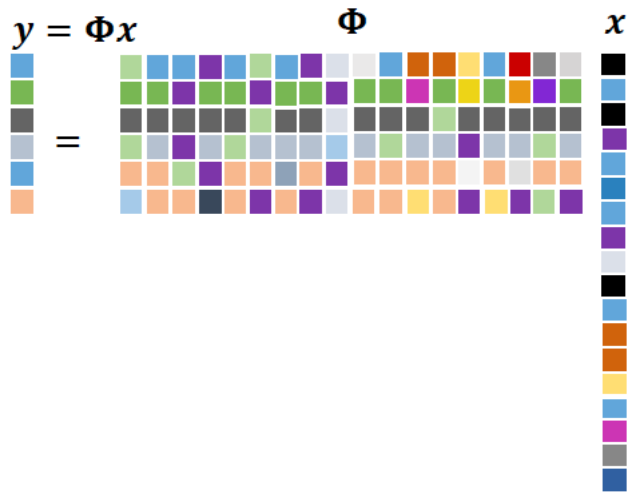


COMPRESSIVE SENSING AND ENCRYPTION

Input signal: $\mathbf{x} \in \mathbf{R}^N$

Sensing matrix: $\Phi \in \mathbf{R}^{M \times N}, M < N$

Measurement vector: $\mathbf{y} = \Phi \mathbf{x}$



Requirements for signal recovery

- Signal sparsity
- Design of the sensing matrix

Information secrecy

- Plaintext: \mathbf{x}
- Ciphertext: \mathbf{y}
- Gaussian one-time sensing

Computational secrecy

$$\mathbf{y}_i = \Phi_i \mathbf{x}_i$$

Perfect secrecy

Mutual information:

$$I(\mathbf{x}_i, \mathbf{y}_i) = I(E_{\mathbf{x}_i}, E_{\mathbf{y}_i})$$

$E_{\mathbf{x}_i}$ is energy of \mathbf{x}_i and $E_{\mathbf{y}_i}$ is energy of \mathbf{y}_i .

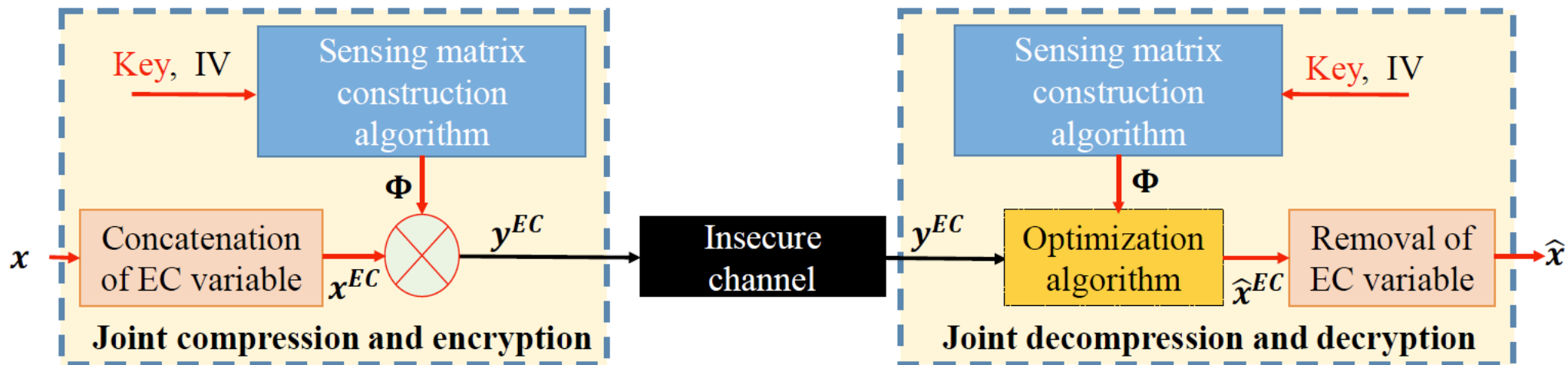
ENERGY CONCEALMENT ENCRYPTION SCHEME

Objective: To tackle the weaknesses of the state-of-the-art CS-based encryption systems

- CS ciphertext leaks energy
- Plaintext are correlated

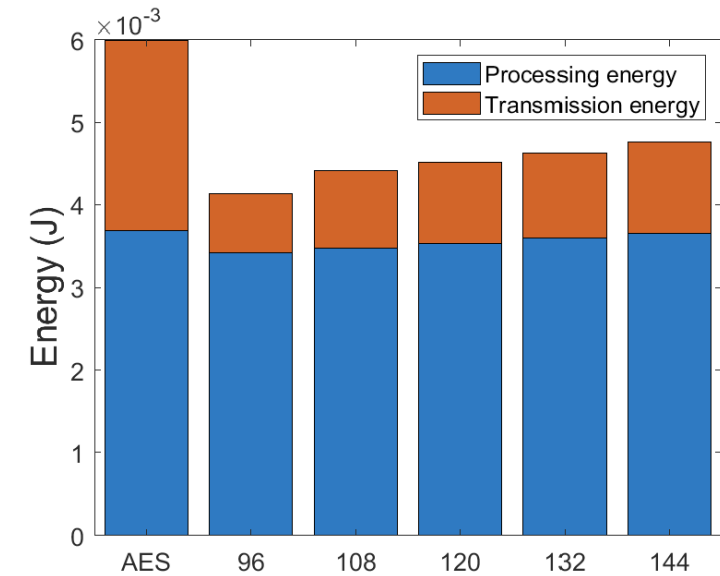
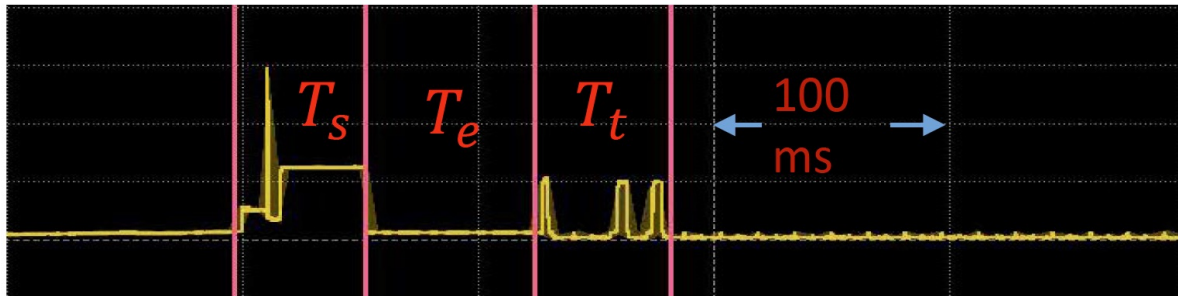
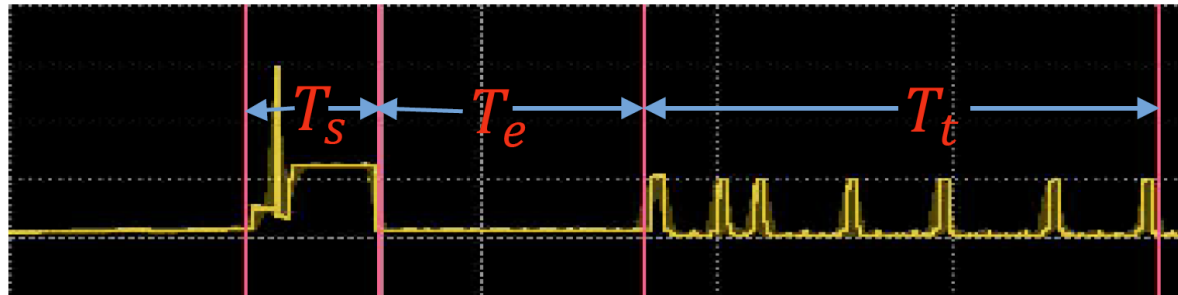
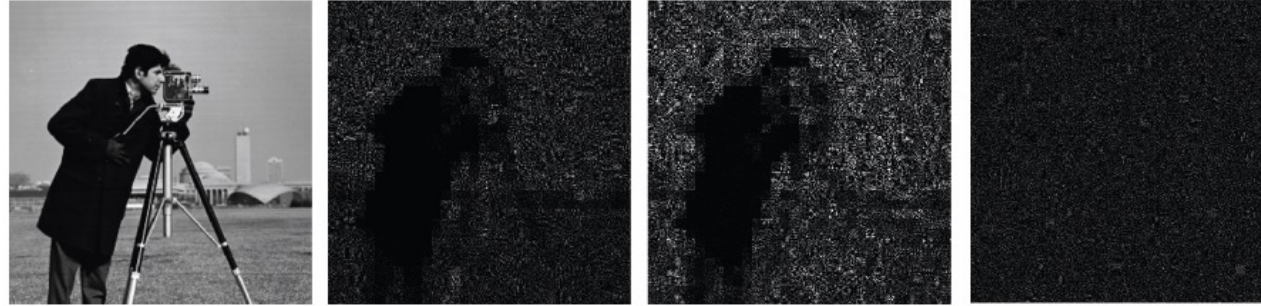
For data $\mathbf{x} = [x_1, x_2, \dots, x_{N-1}]^T$, construct an energy concealment $\mathbf{x}^{EC} = [c, x_1, x_2, \dots, x_{N-1}]^T$

Using Compressive Sensing to compress data $\mathbf{x}^{EC} \in R^N$ to $\mathbf{y}^{EC} \in R^M$, with $CR = M/N$



G. Kuldeep and Q. Zhang, "Design Prototype and Security Analysis of a Lightweight Joint Compression and Encryption Scheme for Resource-Constrained IoT Devices," in IEEE Internet of Things Journal, 2022.

EC: PERFORMANCE EVALUATION



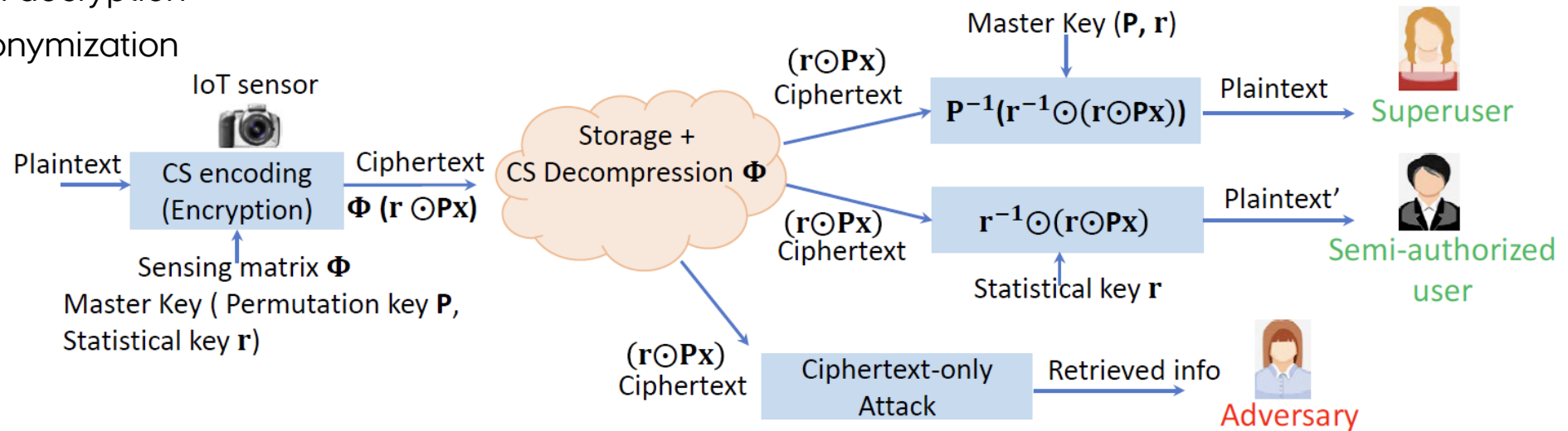
MULTI-CLASS PRIVACY PERSEVERING CLOUD COMPUTING (MPCC)

Objectives:

- Multi-class encryption
- Privacy preserving computation intensive signal recovery at cloud
- Joint compression and information secrecy

MPCC applications:

- Statistical decryption
- Data anonymization



G. Kuldeep and Q. Zhang, "Multi-class Privacy-preserving Cloud Computing based on Compressive Sensing for IoT," Elsevier Journal of Information Security and Applications, 2022.

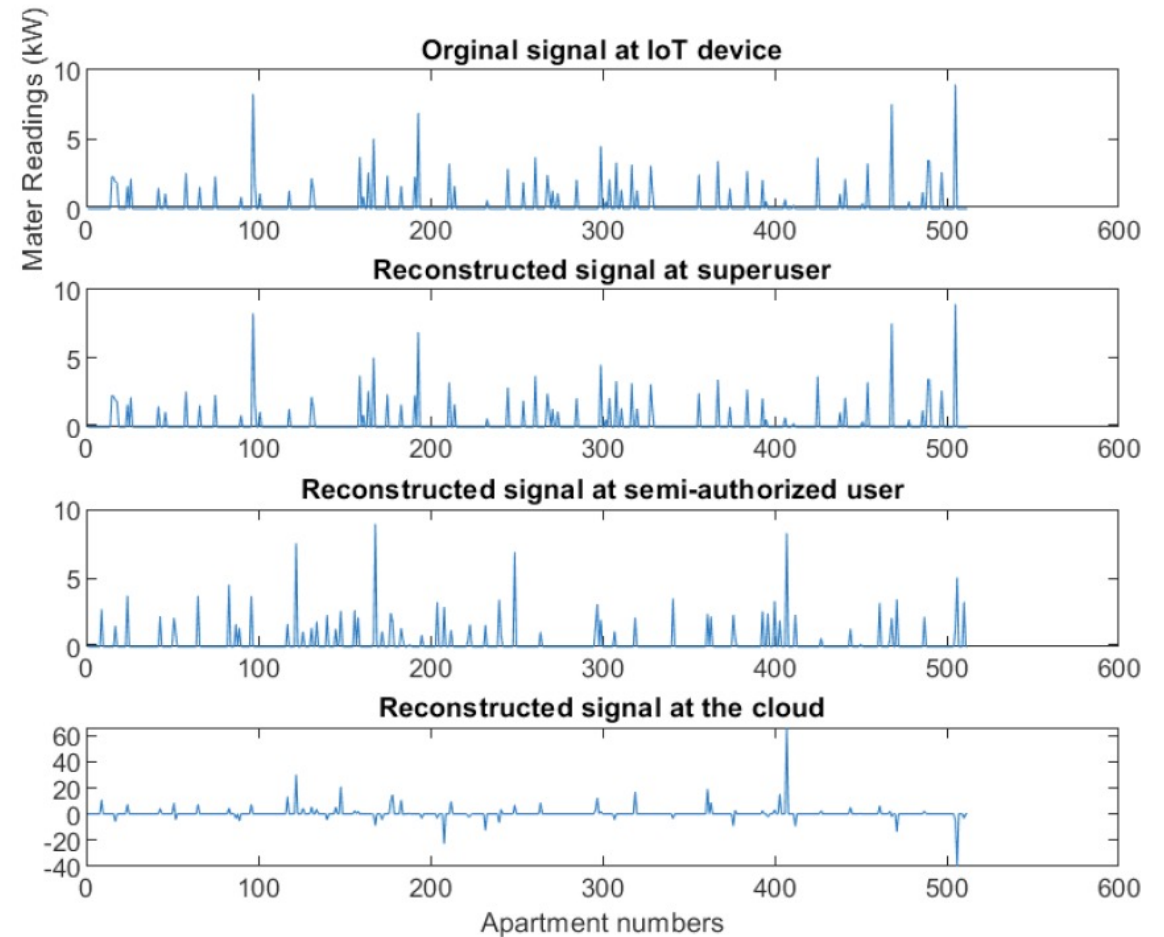
MPCC: STATISTICAL DECRYPTION-1

Access point: joint compression and encryption of smart meter readings

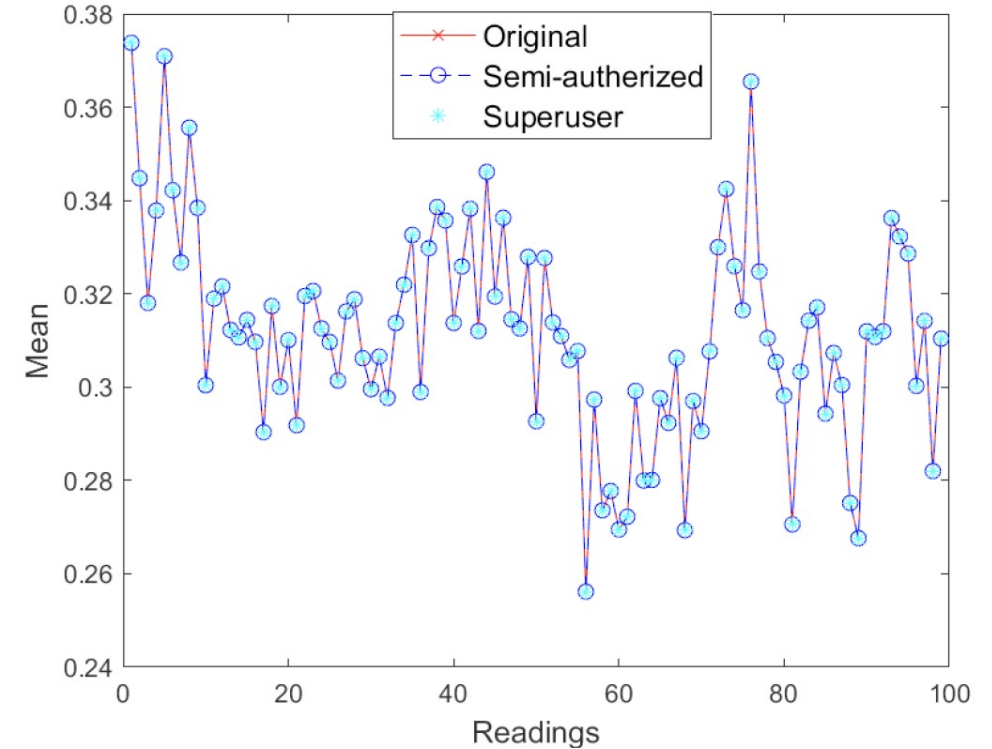
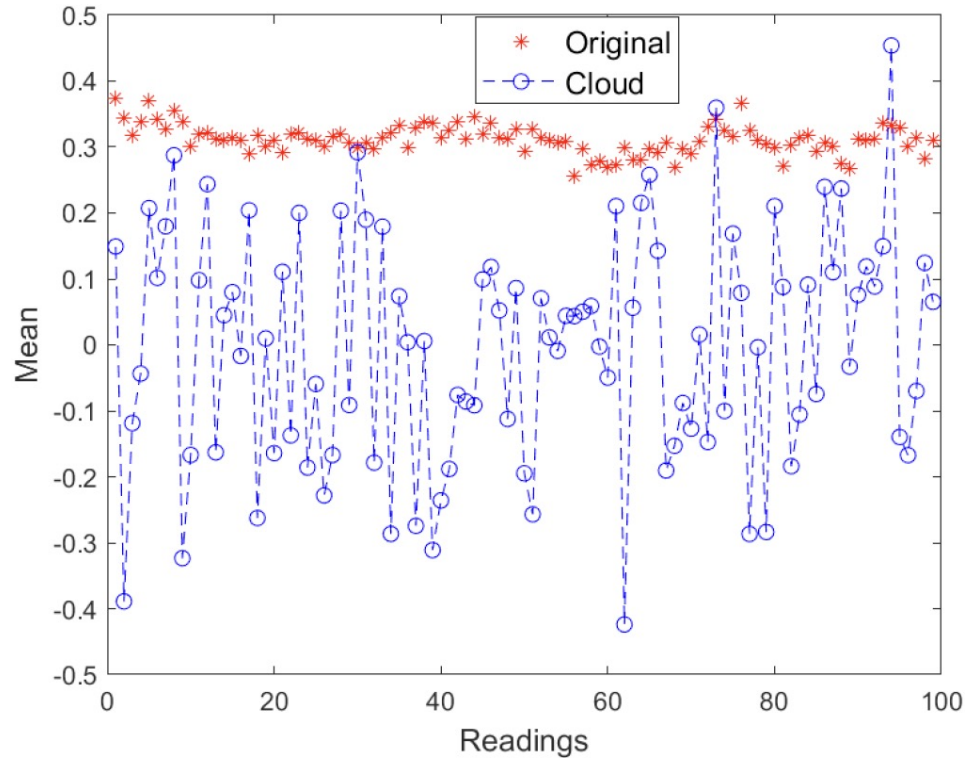
Cloud: storage and decompression

Superuser: exact meter readings

Semi-authorized user: only statistical information



MPCCC: STATISTICAL DECRYPTION-2



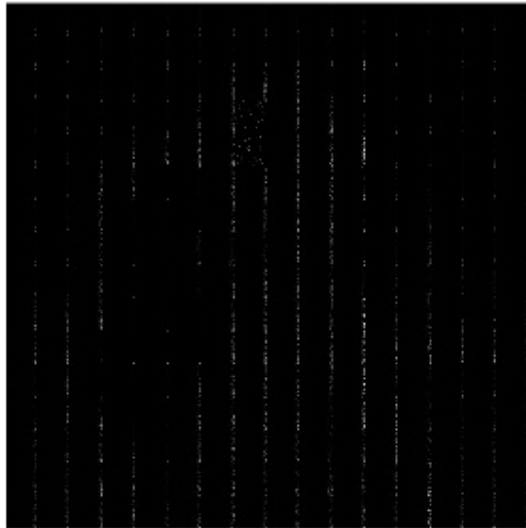
MPCCC: DATA ANONYMIZATION

IoT device: joint compression and encryption of images

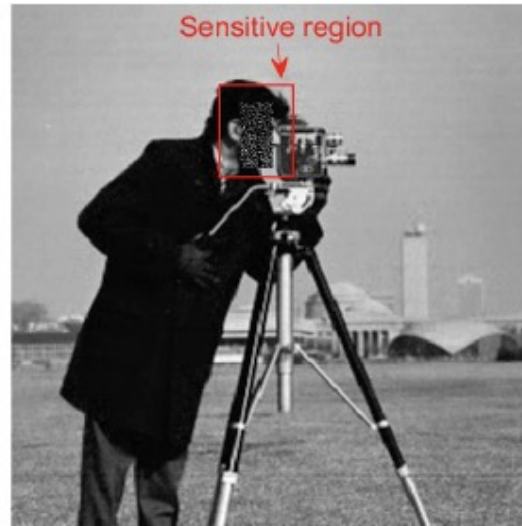
Cloud: storage and decompression

Superuser: complete image

Semi-authorized user: non-sensitive part of the image



Reconstructed image at the cloud

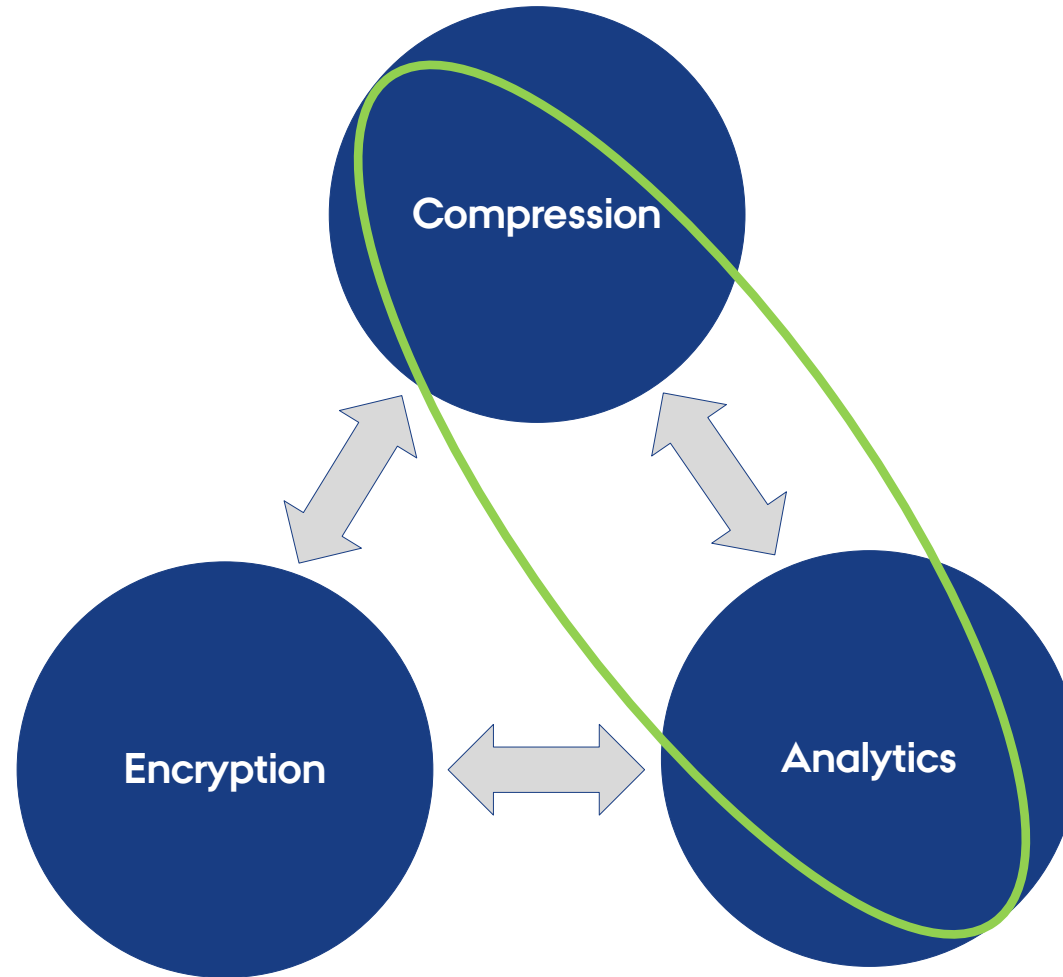


At semi-authorized user

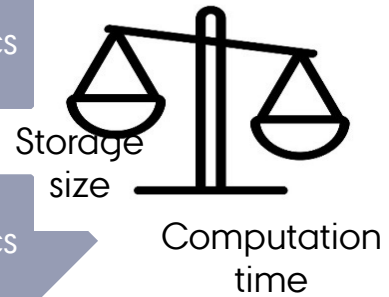
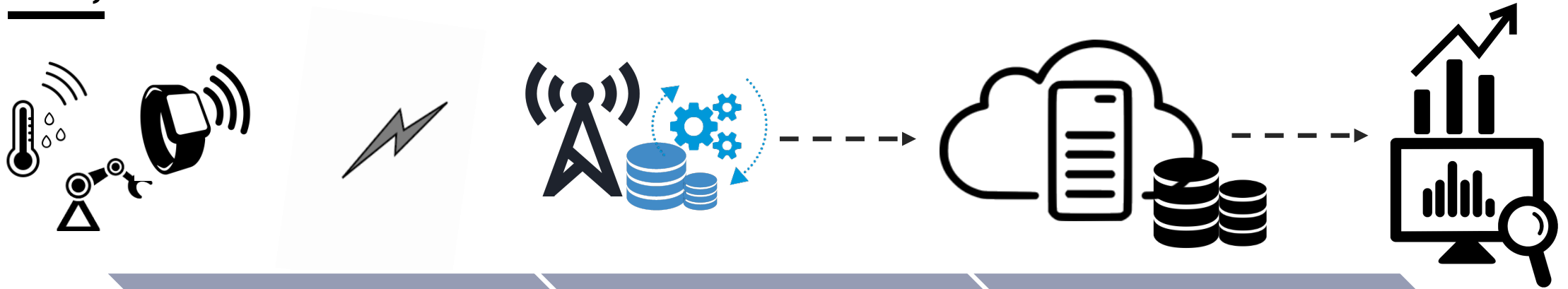


At super-user

JOINT COMMUNICATION AND COMPUTING



IOT FRAMEWORK FOR DATA ACQUISITION, TRANSMISSION, STORE, AND ANALYTICS



FOR IOT DATA ACQUISITION, TRANSMISSION, STORE, AND ANALYTICS

Goal: to create a sustainable IoT solution through a holistic, end-to-end framework to address the challenges,

- Minimize IoT data traffic and storage;
- Save energy in IoT devices, communication and storage infrastructure, as well as in computing;
- Accelerate data analysis, striking a balance between accuracy and efficiency;

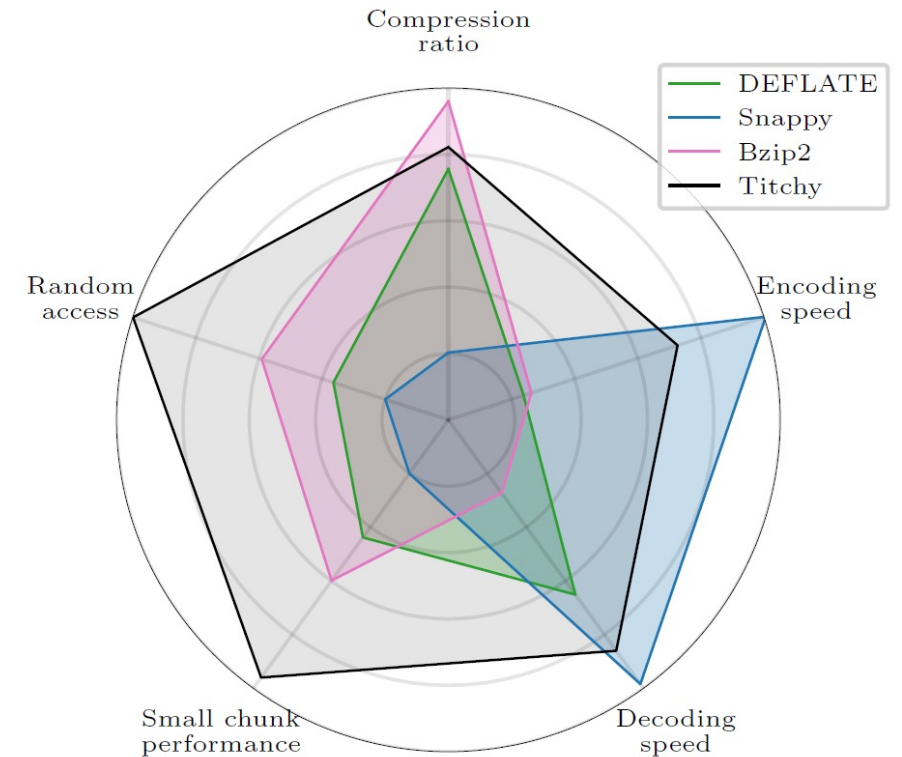
Solution:

- New lossless data compression algorithm, **Generalized Deduplication**, enables
 - random data access;
 - direct data analytics;
 - and could provide an opportunity for privacy-preserving analytics.

GENERALIZED DEDUPLICATION

TITCHY: TIME-SERIES COMPRESSION WITH RANDOM ACCESS

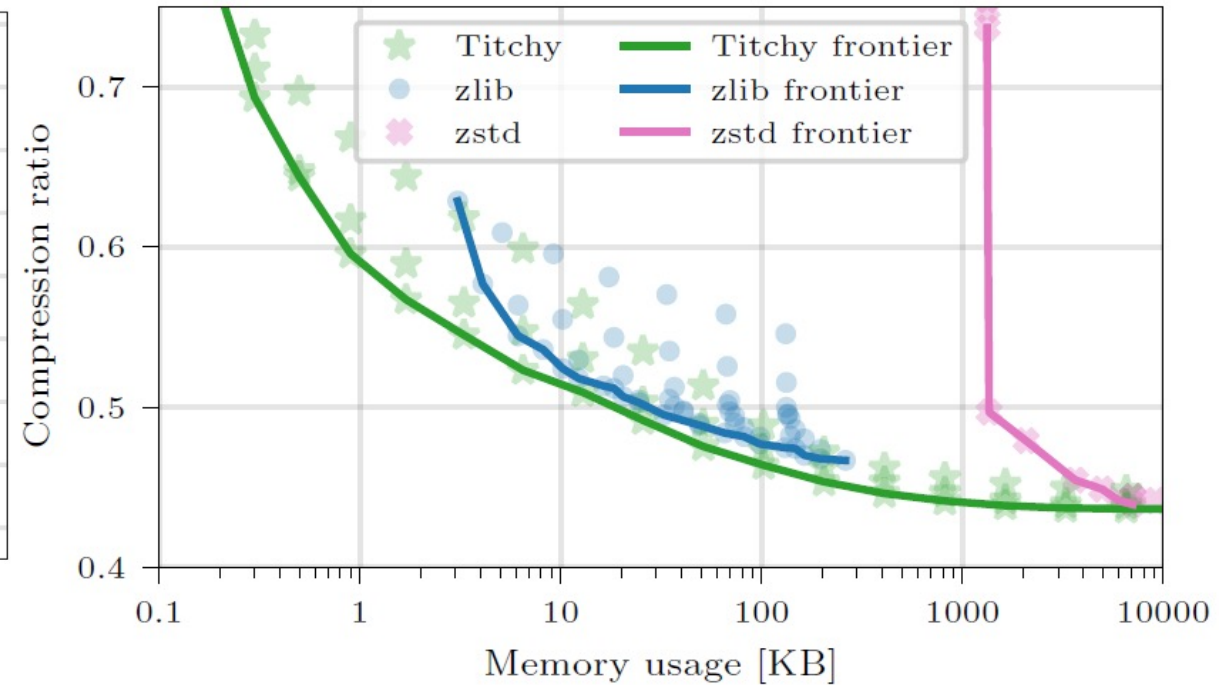
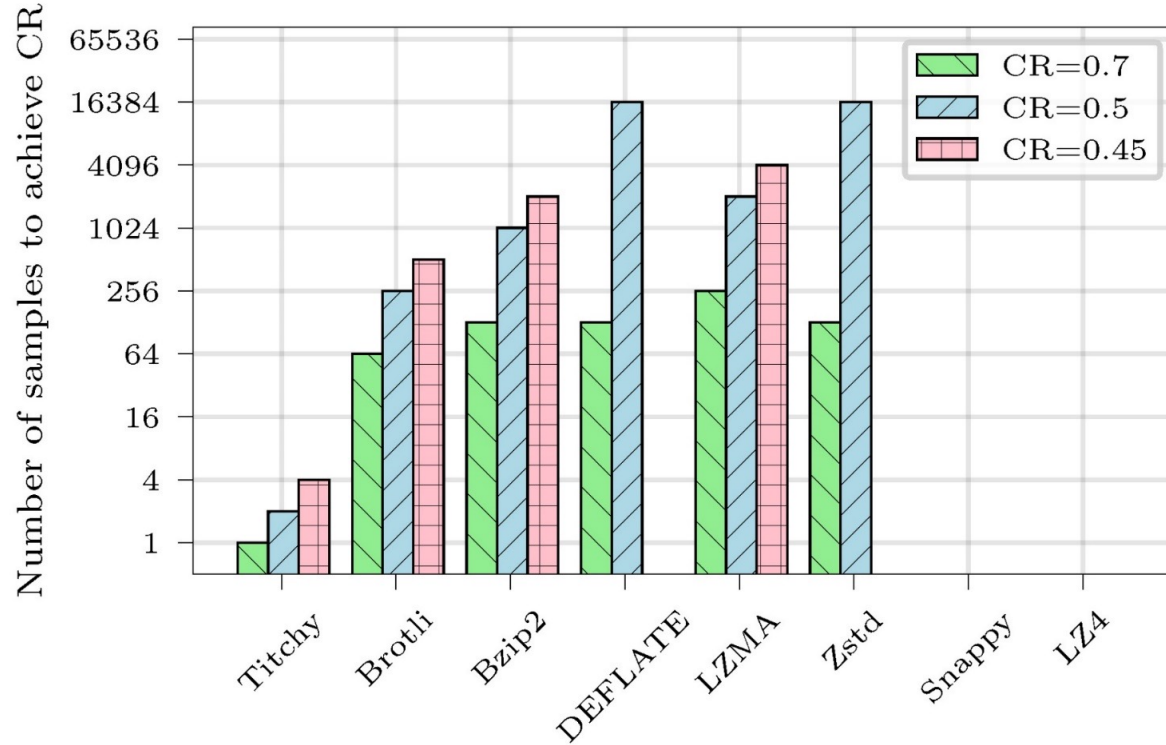
- Data Compression at resource-constrained IoT devices
- To realize end-to-end compression, with good overall performance in
 - Compression ratio
 - Encoding speed
 - Decoding speed
 - **Small chunk performance**
 - **Random access capability**



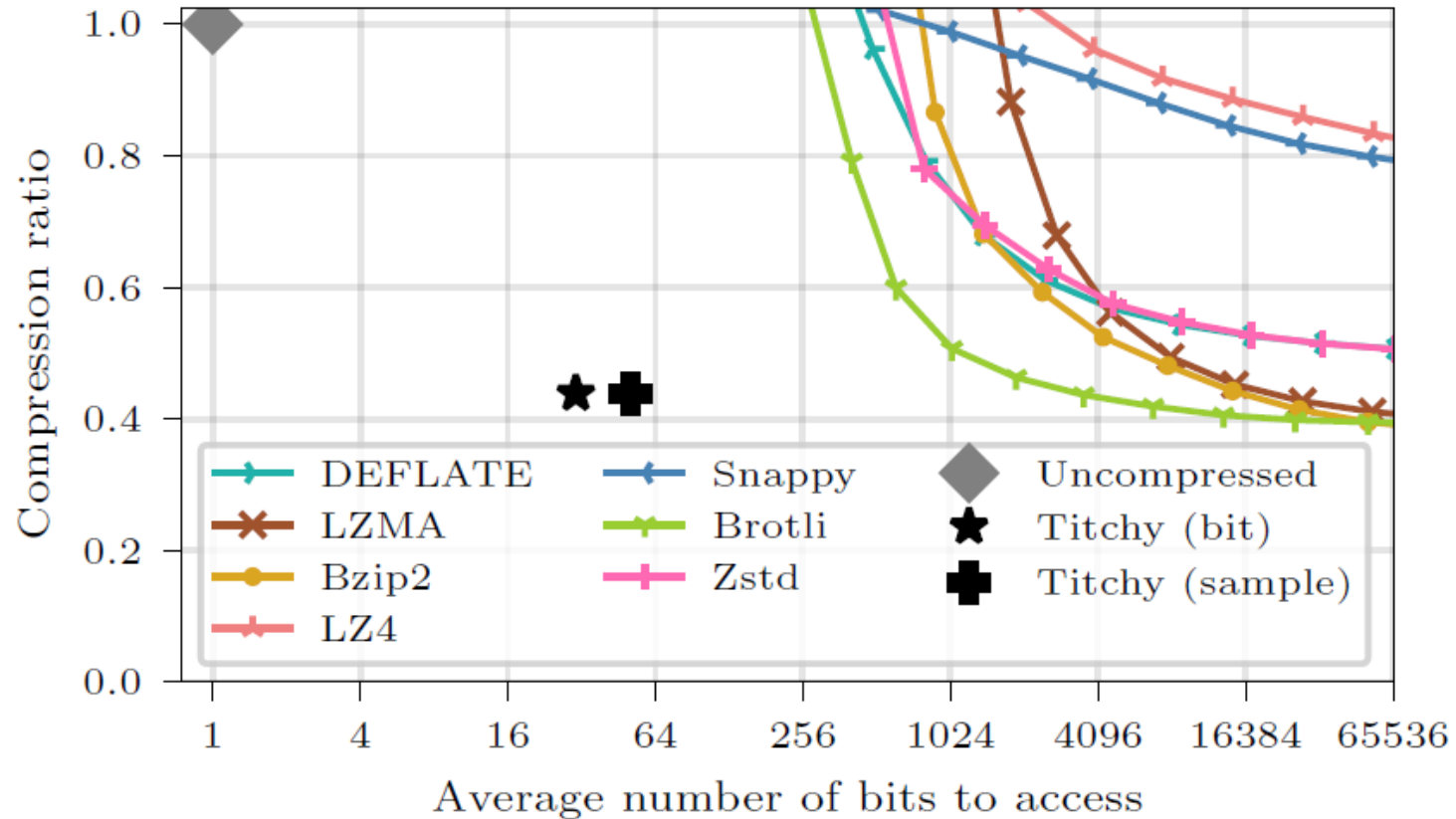
- R. Vestergaard, D. E. Lucani Rötter; Q. Zhang. A Randomly Accessible Lossless Compression Scheme for Time-Series Data. IEEE INFOCOM 2020. s. 2145-2154.

- R. Vestergaard, Q. Zhang, M. Sipos, D. E. Lucani Rötter, "Titchy: Online Time-series Compression with Random Access for the Internet of Things", IEEE Internet of Things Journal, 05.2021.

COMPRESSION RATIO VS. BLOCK SIZE & MEMORY USAGE



RANDOM ACCESS COST



Can we perform data query processing and analytics directly on the GD compressed data?

Bases

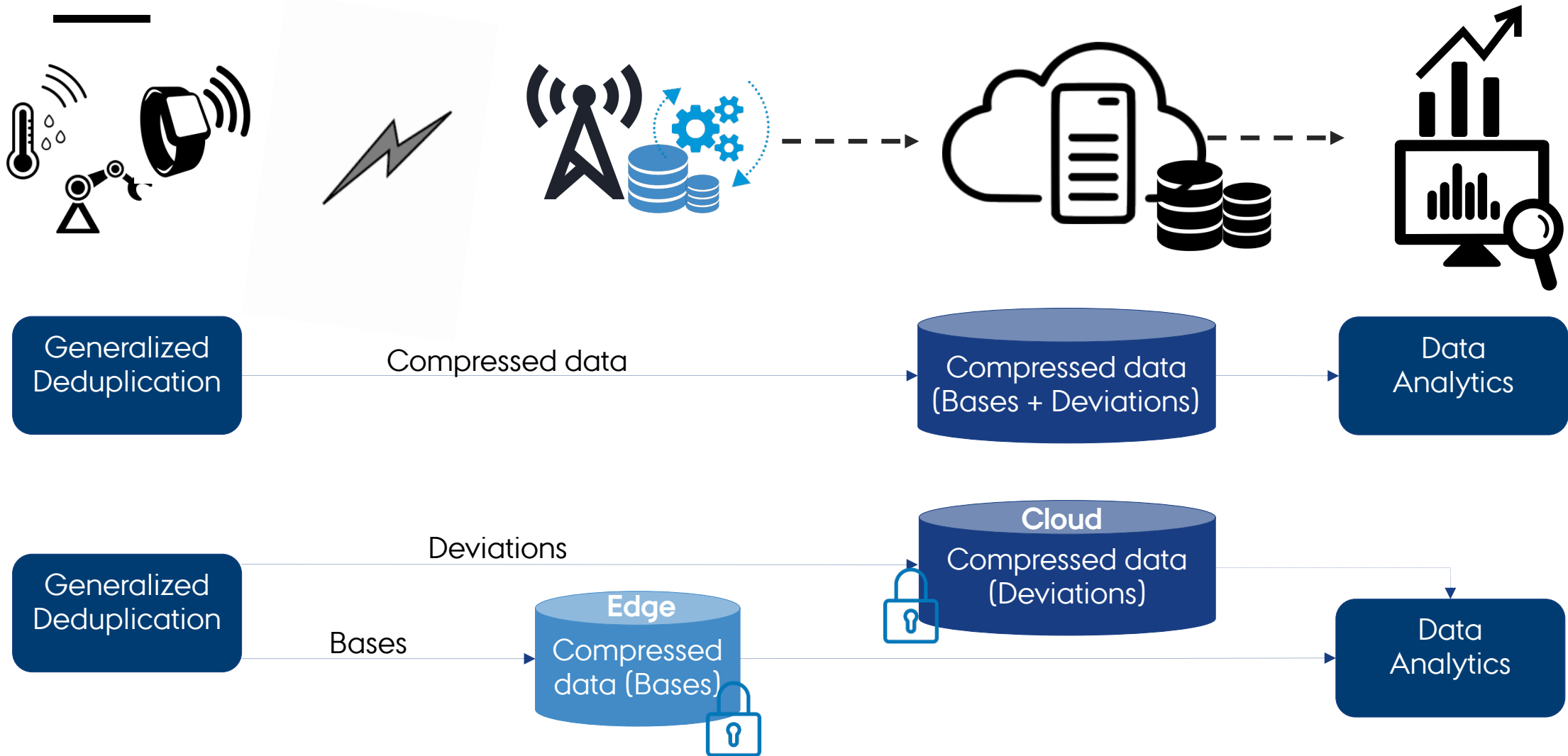
```
1010111011011
0101010010101
001010101001...
```

The bases act like a **summary** of the data.

Deviations

```
101011101101101010100100101
010100101010010101010100100
10101101111111010010111001...
```

A JOINT COMMUNICATION AND COMPUTING IOT FRAMEWORK



CLUSTERING ON GD COMPRESSED DATA

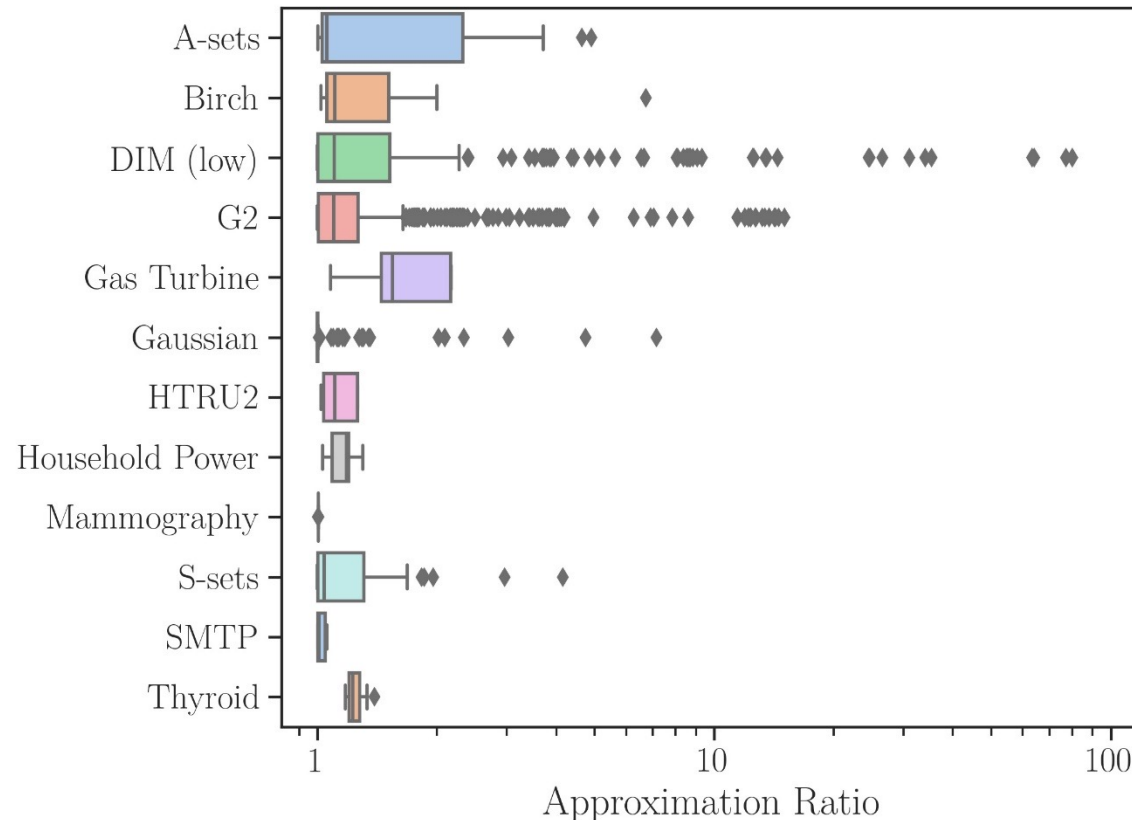
Dataset Group	Description	# Datasets	n	d	k	Mean Size (kB)	Type	Comp. Configs	
								Total	Mean
SYNTHETIC	A-sets [40]	3	3,000–7,500	2	20–50	42.1	int	24	8
	Birch [40]	2	100,000	2	100	800.1	int	21	11
	DIM (low) [40], [41]	14	1,351–10,126	2–15	9	239.1	int	292	21
	G2 [40]	60	2,048	2–64	2	172.2	int	924	16
	S-sets [40]	4	5,000	2	15	40.1	int	32	8
	Gaussian [28]	10	100,000	2	5	800.1	float	100	10
REAL	Gas turbine [42]	1	36,733	11	5	1,616.4	float	22	22
	Power consumption [43]	5	10,000–400,000	7	4	4,213.2	float	35	7
	HTRU2 [44]	1	17,898	8	5	572.9	float	10	10
	Mammography [45]		11,183	6	6	268.5	float	10	10
	SMTP [45]		95,156	3	5	1,142.0	float	10	10
	Thyroid [45]		3,772	6	7	90.7	float	10	10
Total		103						1,490	14.5

A. Hurst, D. E. Lucani, I. Assent and Q. Zhang, “GLEAN: Generalized Deduplication Enabled Approximate Edge Analytics”, IEEE Internet of Things Journal, 2022

DATA CLUSTERING QUALITY ON COMPRESSED DATA WITH DIFFERENT GD CONFIGURATIONS

Only access the bases of GD compressed data

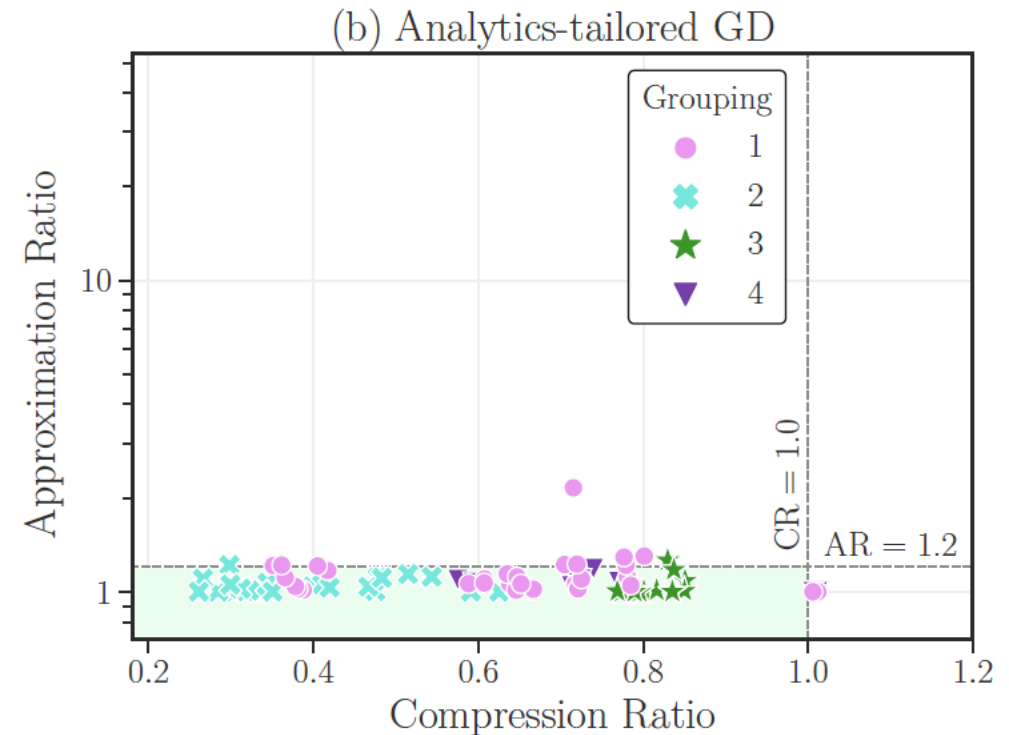
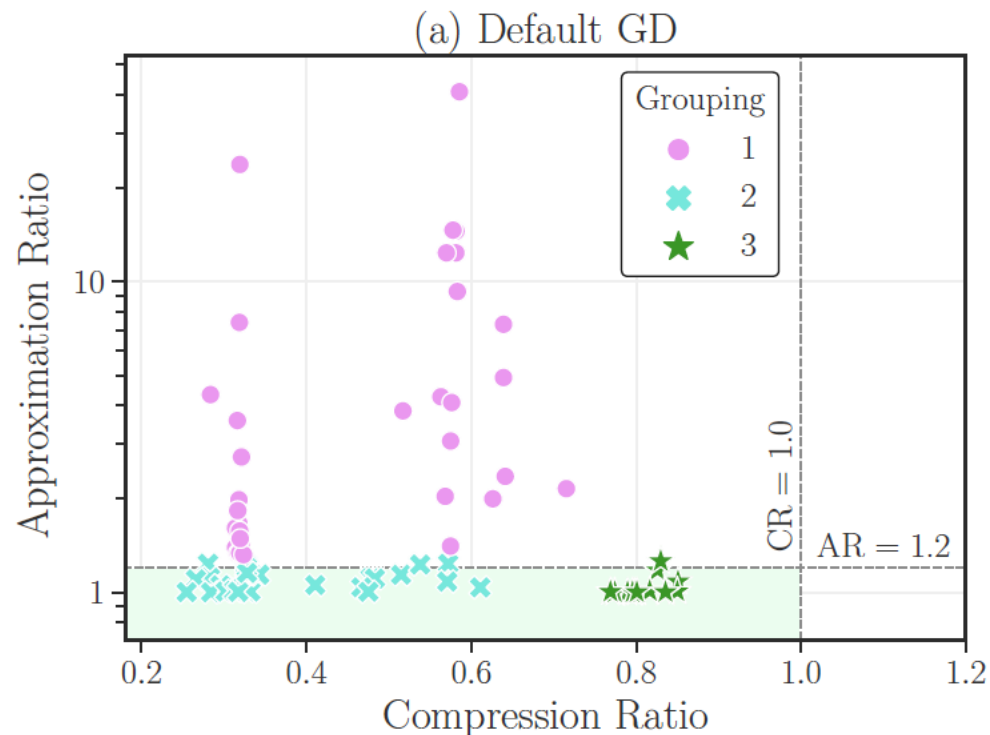
Approximation Ratio = sum of squared errors (basis) / sum of squared errors (x)



OPTIMIZE GD TO IMPROVE CLUSTERING QUALITY

Optimizing GD

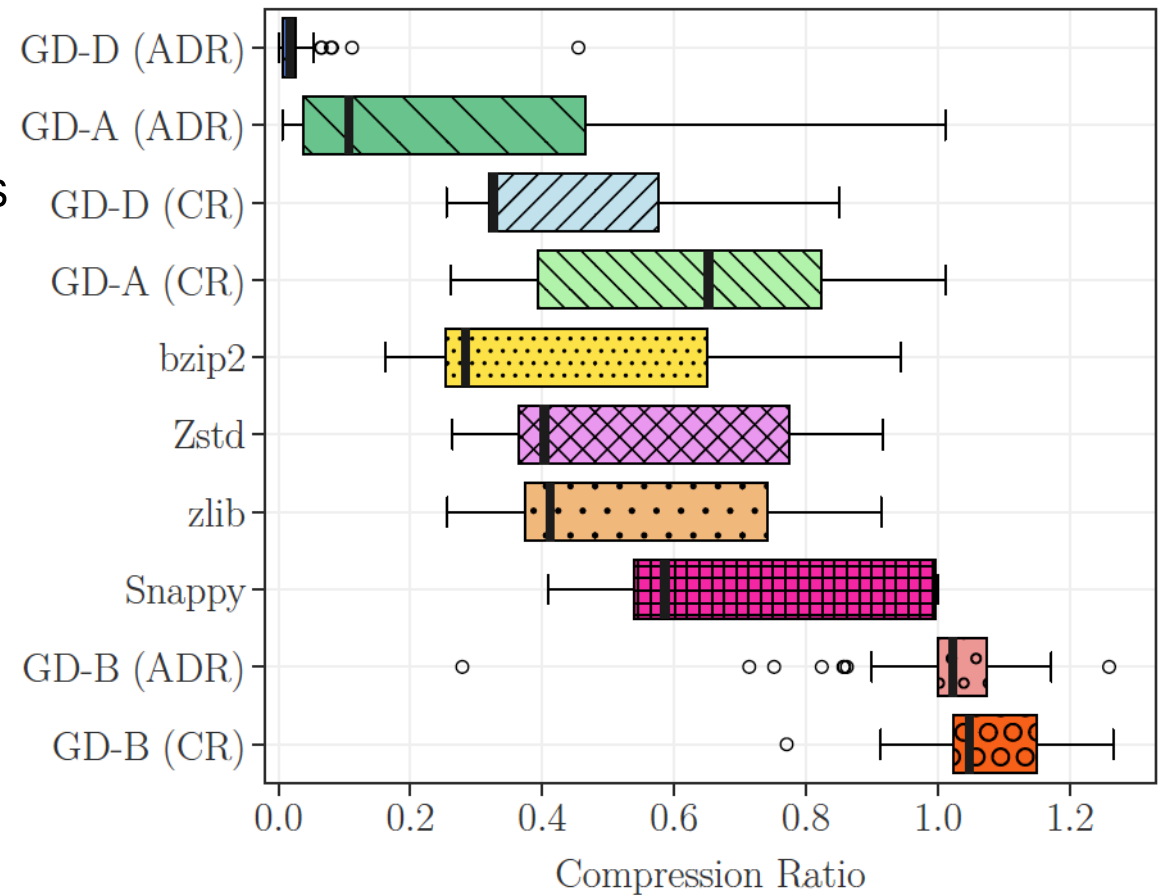
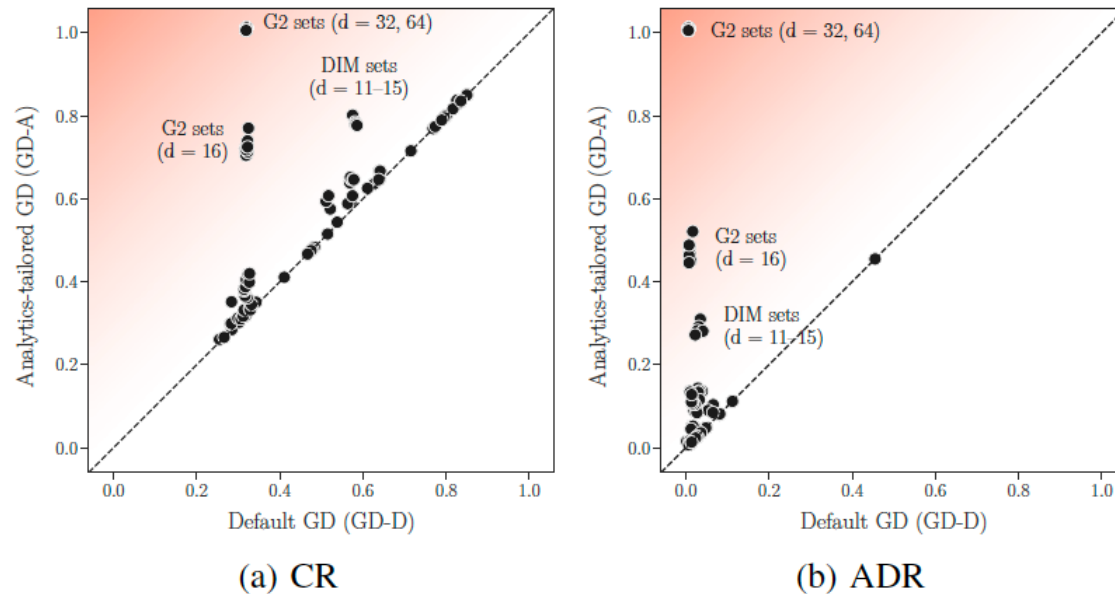
- Design a heuristic approach
- Improve data clustering quality at the expense of compression ratio



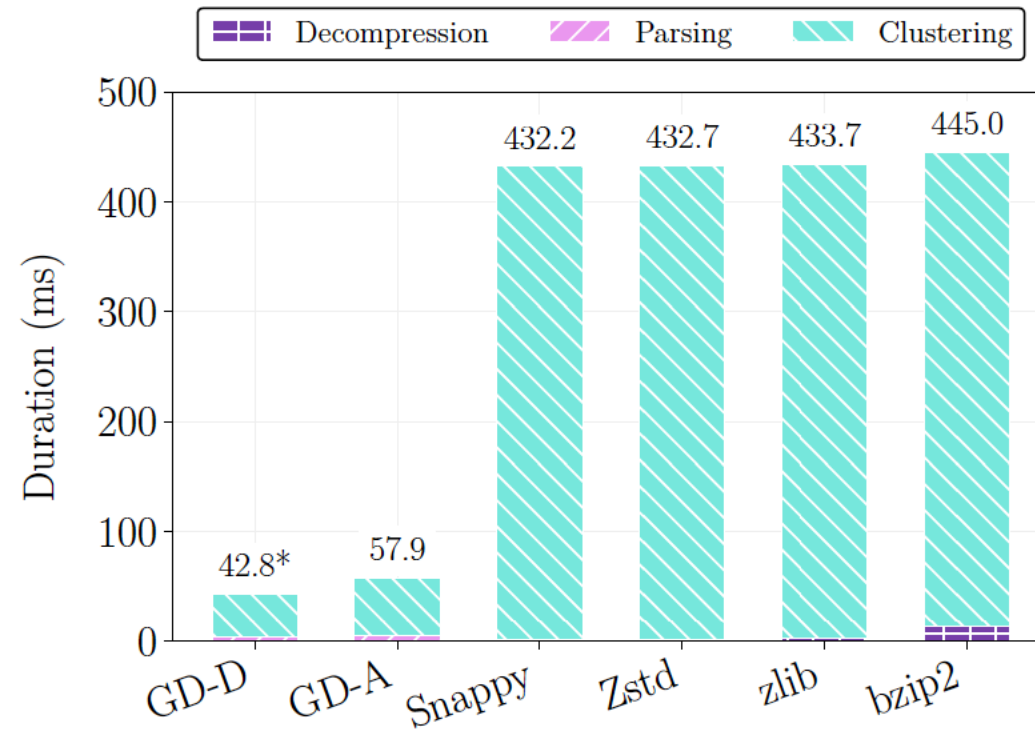
ANALYTICS DATA RATIO AND COMPRESSION RATIO

Compression ratio and Analytics data ratio

- GD vs. SoTA universal data compressors



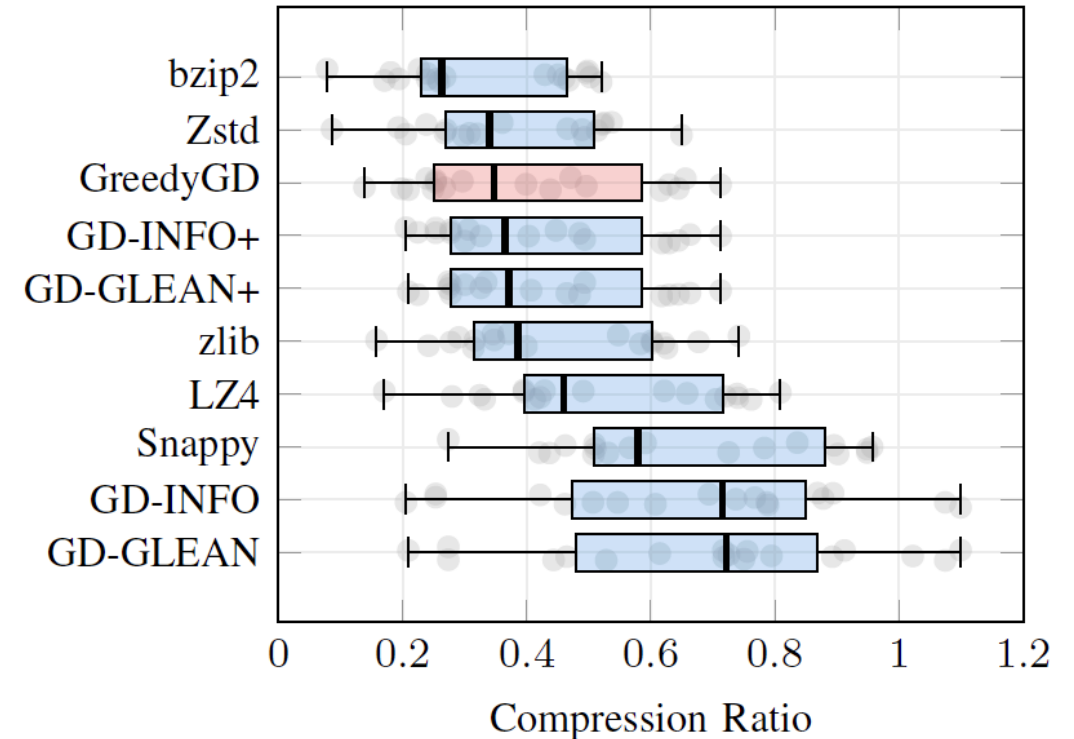
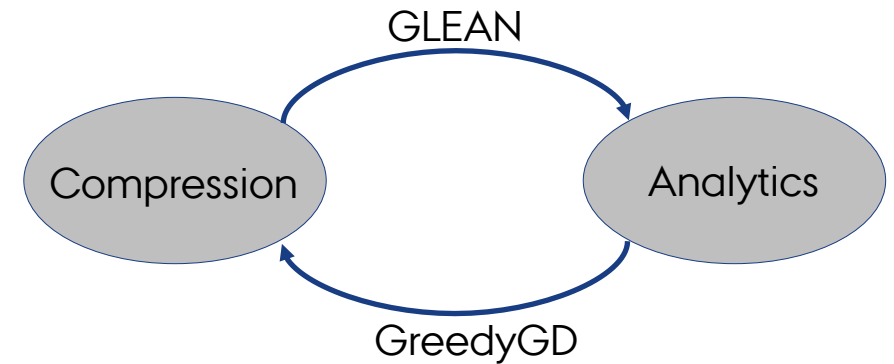
RUNTIME COMPARISON



8-10X

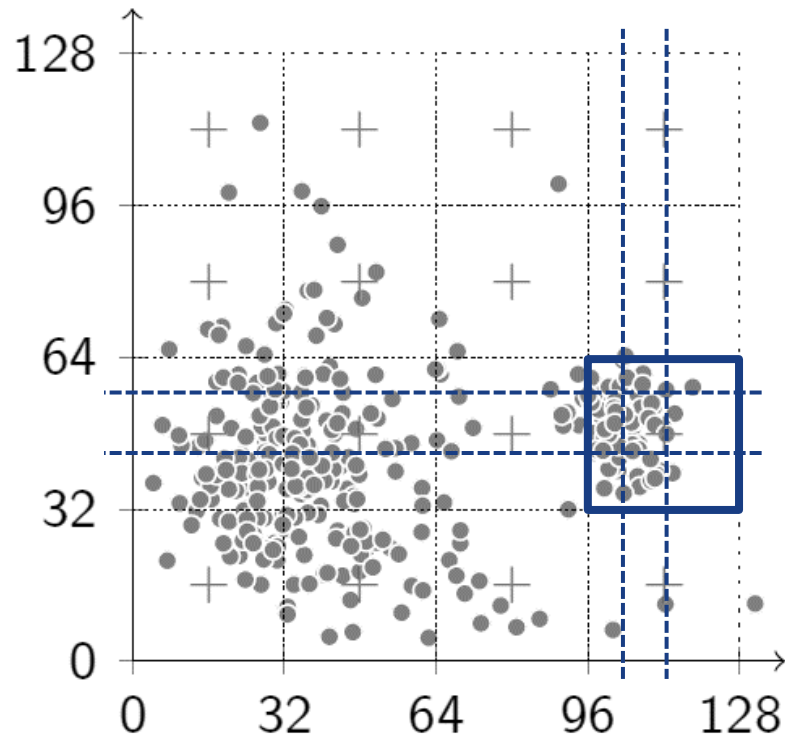
GREEDYGD

Dataset	Precision	n	d	Size (kB)
Aarhus weather [19]	single	26,387	4	422
Aarhus pollution #172156 [21]	single	17,568	5	351
Aarhus pollution #204273 [21]	single	17,568	5	351
Chicago beach water [22]	single	39,829	5	797
Chicago beach water (w/depth) [22]	single	10,034	6	241
Chicago beach weather (float) [23]	single	86,694	9	3,121
Chicago beach weather (integer) [23]	single	86,763	5	1,735
Chicago taxi trips [24]	double	3,466,498	10	277,320
CMU IMU (accelerometer) [25]	single	134,435	3	1,613
CMU IMU (angular velocity) [25]	single	134,435	3	1,613
CMU IMU (magnetometer) [25]	single	134,435	3	1,613
CMU IMU (position) [25]	single	134,435	4	2,151
CMU IMU (all) [25]	single	134,435	13	6,991
COMBED Mains power [26]	double	82,888	3	995
COMBED UPS power [26]	double	86,199	3	1,035
Melbourne city environment [27]	single	56,570	3	679
Gas turbine emissions [28]	single	36,733	11	1,616
Household power consumption [29]	single	2,049,280	7	57,380



APPROXIMATE QUERY PROCESSING

Compute additional metadata to support analytics: **multi-dimensional histograms**



Query median percentage error
(MEAN queries)

Query aggregated on column...

Query predicate on column...

	1	2	3	4	5	6	7
1	.08	.17	.00	.10	.67	.29	.76
2	.12	.21	.00	.14	.40	.27	.13
3	.27	.14	.00	.25	.99	.56	.32
4	.10	.05	.00	.02	.12	.06	.01
5	.04	.02	.00	.03	.00	.00	.00
6	.04	.06	.00	.05	.00	.00	.00
7	.06	.04	.00	.05	.00	.00	.00

Overall median error: **0.02%**

95th percentile error: **0.54%**

Amount of data accessed: **2.3%**



CONCLUSION AND OUTLOOK

- IoT has many challenges but also opportunities.
- A sustainable IoT system requires a holistic design in data- acquisition, compression, encryption, communication, storage and analytics.
 - To bridge Communication with Computing
- Opportunities
 - Time-critical IoT applications also needs joint design of communication and computing
 - It is worth looking into privacy-preserving analytics on compressed data

THANK YOU



DIGIT
AARHUS UNIVERSITY CENTRE FOR
DIGITALISATION BIG DATA AND DATA ANALYTICS



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