

Final Presentation

NAVISP-EL1-060 / VALLE : NOVEL PRIVACY
PRESERVING PNT PROCESSING TECHNIQUES



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Project Introduction

VALLE Team

Prime Contractor



Project Management
Lead Privacy Enhancement Activities

Subcontractors



Lead Navigation Activities



Lead Red Team Analysis

Scope and Objectives

- **The main objectives of VALLE:**

- Identify, define, and consolidate a set of use cases for privacy-preserving positioning solutions or services based on sharing and processing user PNT data.
- Define and develop multiple privacy-preserving PNT processing concepts based on the sharing and processing of different types of user PNT data.
- Design and implement a concept demonstrator to verify and validate the proposed concepts.

Use Cases Trade-Off and Selection

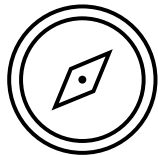
Privacy vs functionality

- Working with data imposes a trade-off between service functionality and data privacy. Privacy may need to be sacrificed to enable certain features and vice versa.

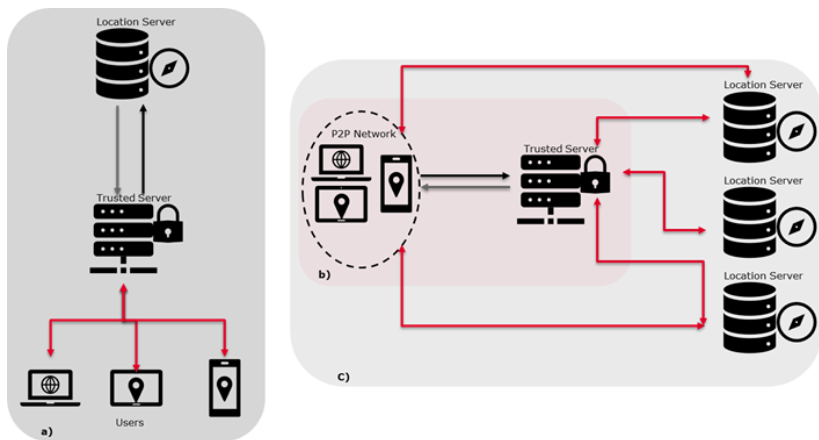


- Novel privacy enhancement techniques, such as the ones proposed in VALLE, promise to overcome this trade-off by allowing calculations without disclosing private data.

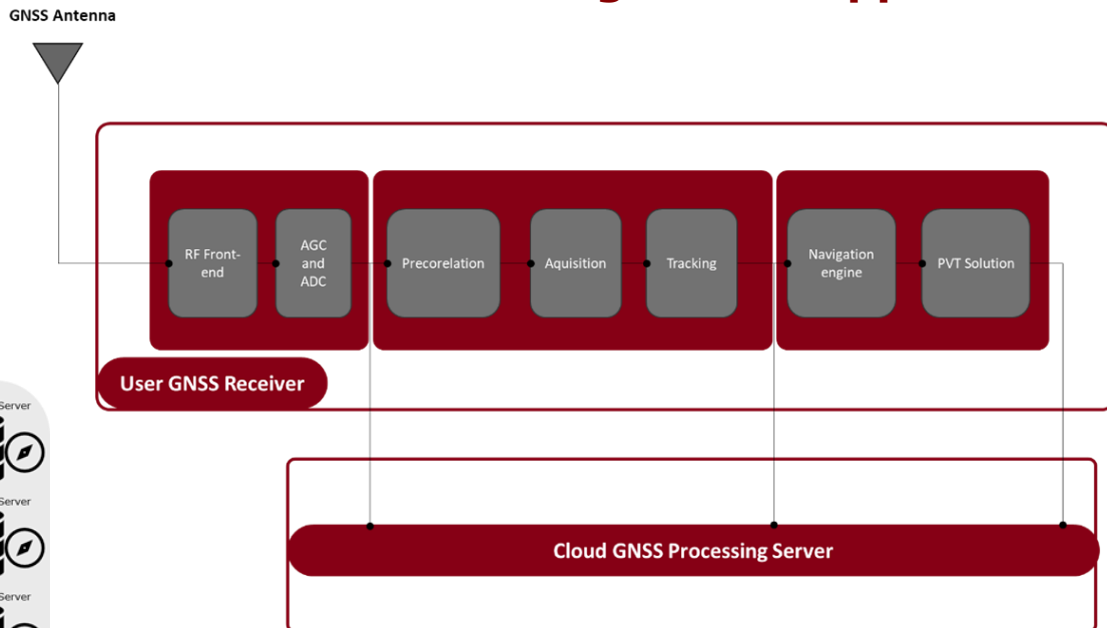
System Architectures



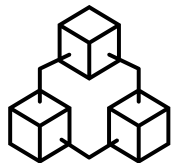
LBS Architectures



Cloud GNSS Processing Server Approach



Privacy Enhancement Techniques Selection



Federated
Learning

Differential
Privacy

Secure Multi
Party
Computation

Anonymization

Homomorphic
Encryption

Zero-
Knowledge
Proof

Trusted
Environment
Execution

Use Cases Final Selection

			Relevance	Complexity (PNT)	Complexity (PET)	Aggregate Score	Data type	PET Technique
✓	1	Secure collaborative positioning (Anonymisation)	High (3)	Low (3)	Low (3)	High (3.0)	Observables	Anonymisation
✓	2	Crowd management applications	High (3)	Low (3)	Medium (2)	High (2.75)	PVT	SMPC
	3	Tracking applications for children	Medium (2)	Low (3)	Low (3)	High (2.5)	PVT	Plain Encryption
✓	4	Cloud processing of data for correlation	High (3)	Medium (2)	High (1)	High (2.25)	IQ Samples	Partial HE
	5	Location for Regulatory applications (digital tachograph or road tolling)	High (3)	Medium (2)	High (1)	High (2.25)	Observables/ PVT	ZKP, SMPC or HE
	6	Geo-fencing	High (3)	Medium (2)	High (1)	High (2.25)	PVT	FHE
	7	Secure collaborative positioning (FHE)	High (3)	Medium (2)	High (1)	High (2.25)	Observables	FHE

Relevance Scoring



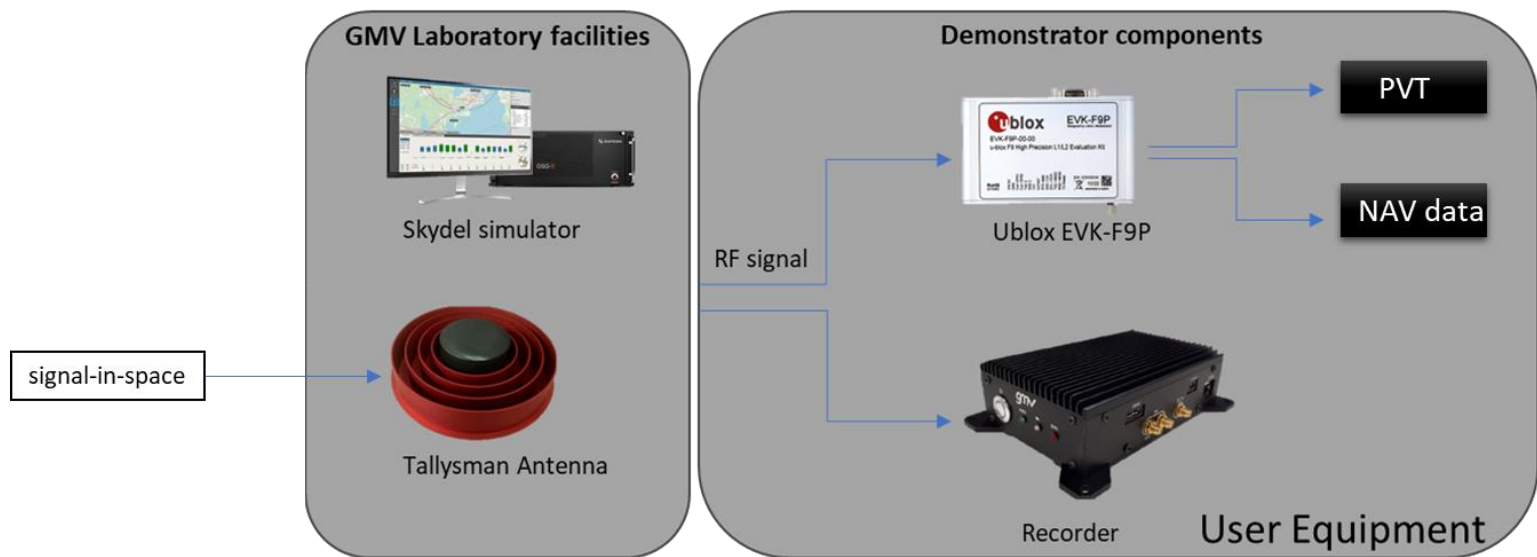
Complexity Scoring



$$\text{Final score} = ((\text{Relevance}) + ((\text{Complexity PNT} + \text{Complexity PET}) / 2)) / 2$$

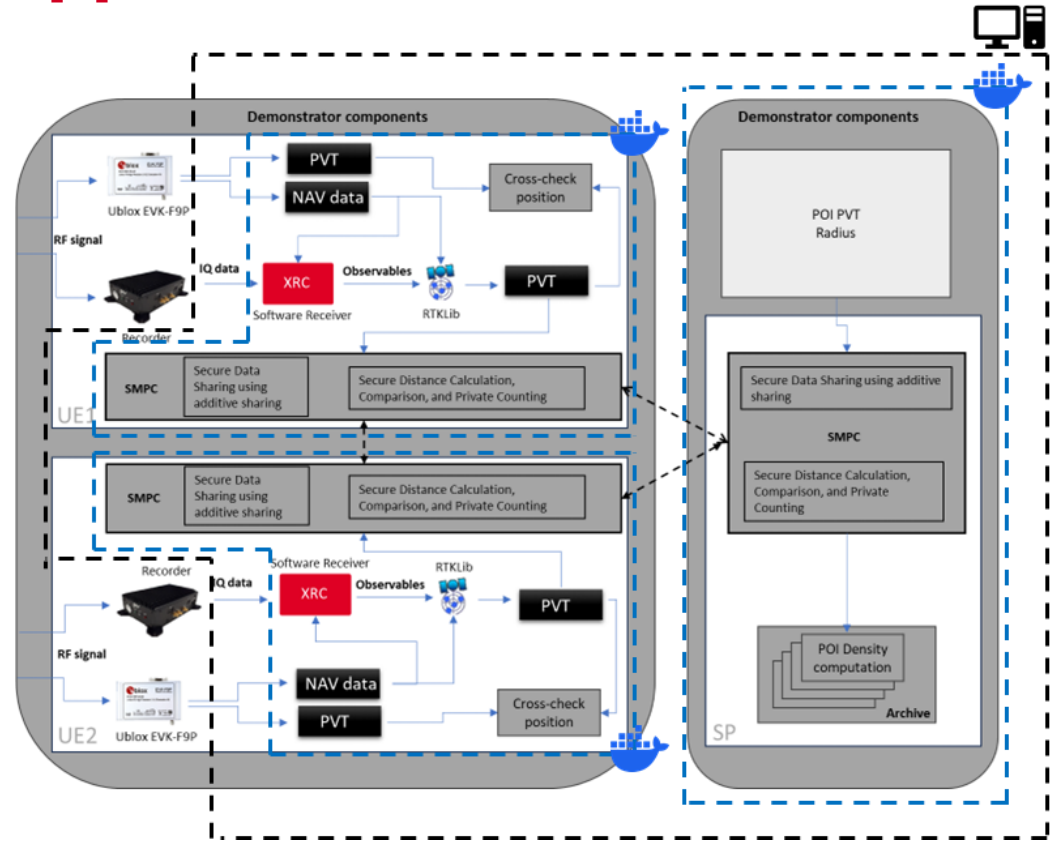
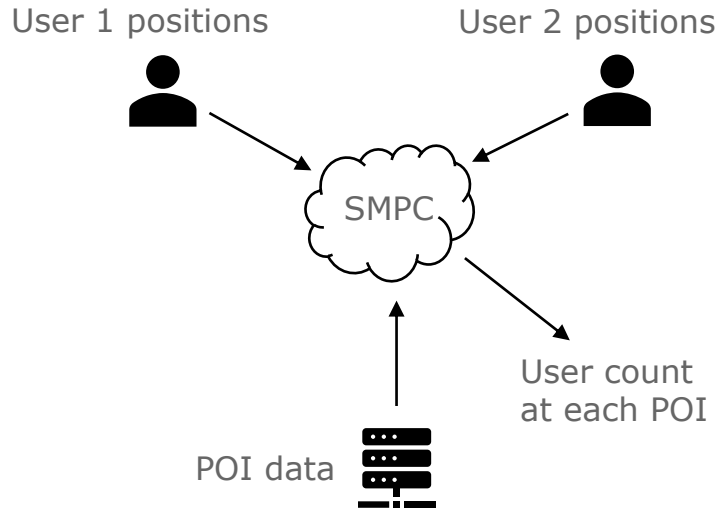
Demonstrator Design and Implementation

Common workflow between use cases



Crowd management applications

SMPC - Secure Multi Party Computation



Secure collaborative positioning

Anonymization

Remove or modify identifying data

UE 1



Dani
(X,Y,Z)

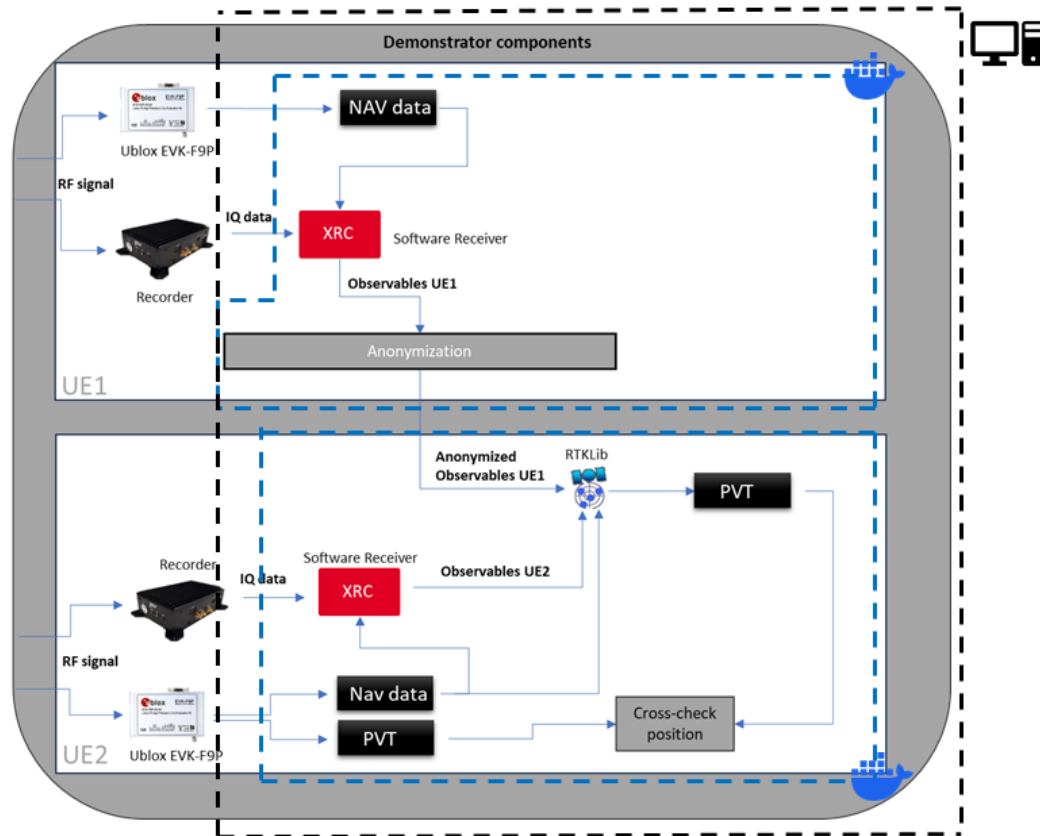
Anonymize headers

UE 2



(X,Y,Z)

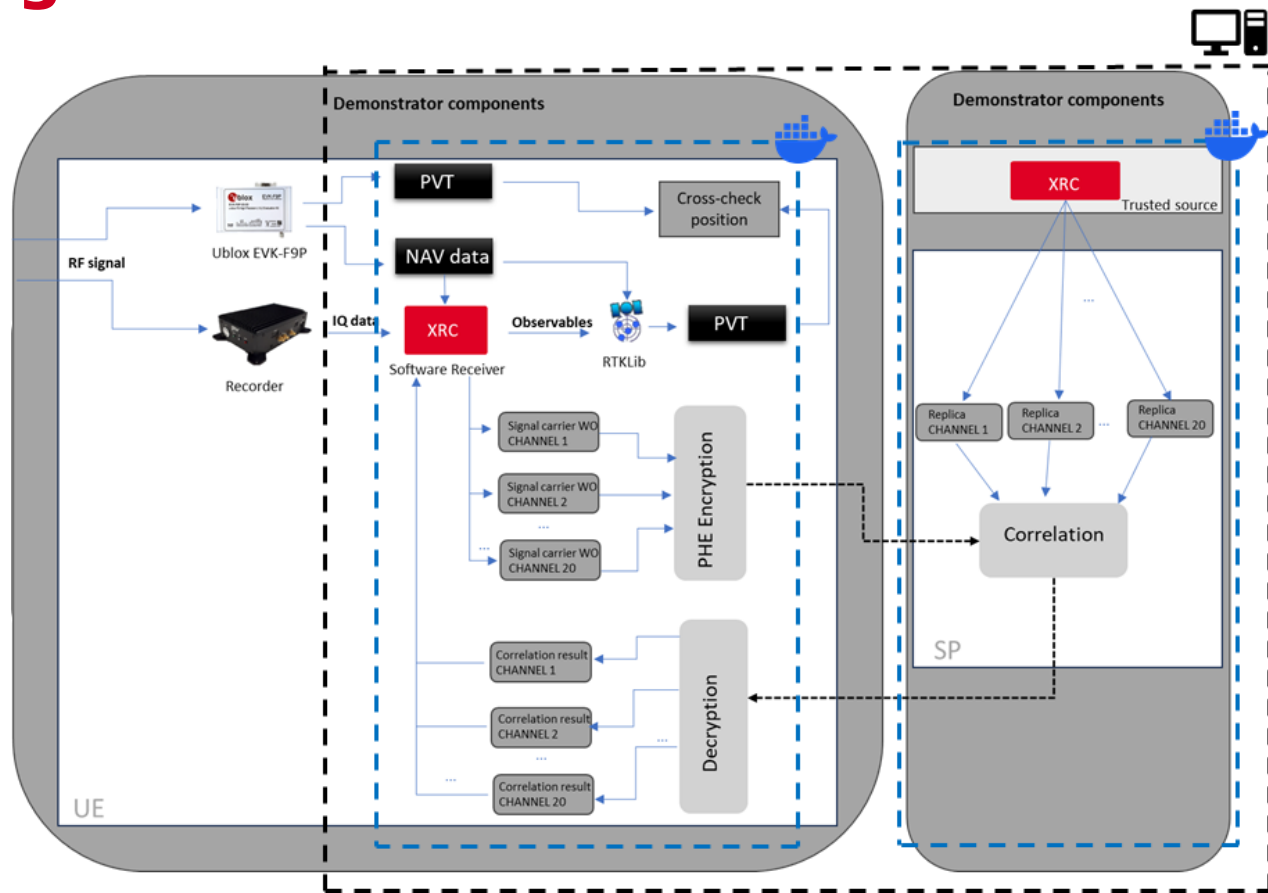
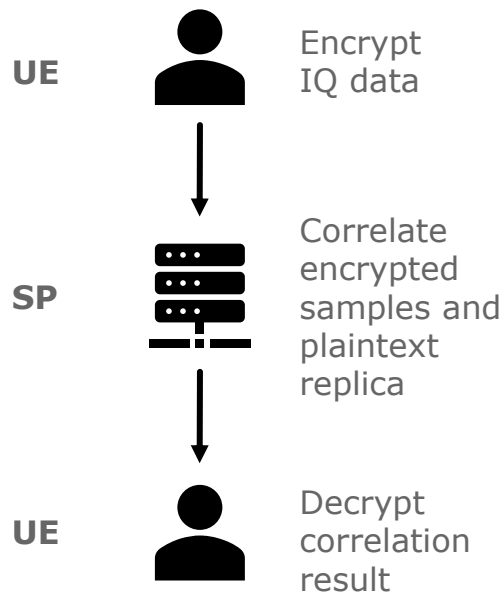
Improve positioning
accuracy



Cloud processing of data for correlation

Partial Homomorphic Encryption

$$\text{enc}(a) \oplus \text{enc}(b) = \text{enc}(a + b)$$



Performance and Red Team Analysis

Computational Performance Analysis

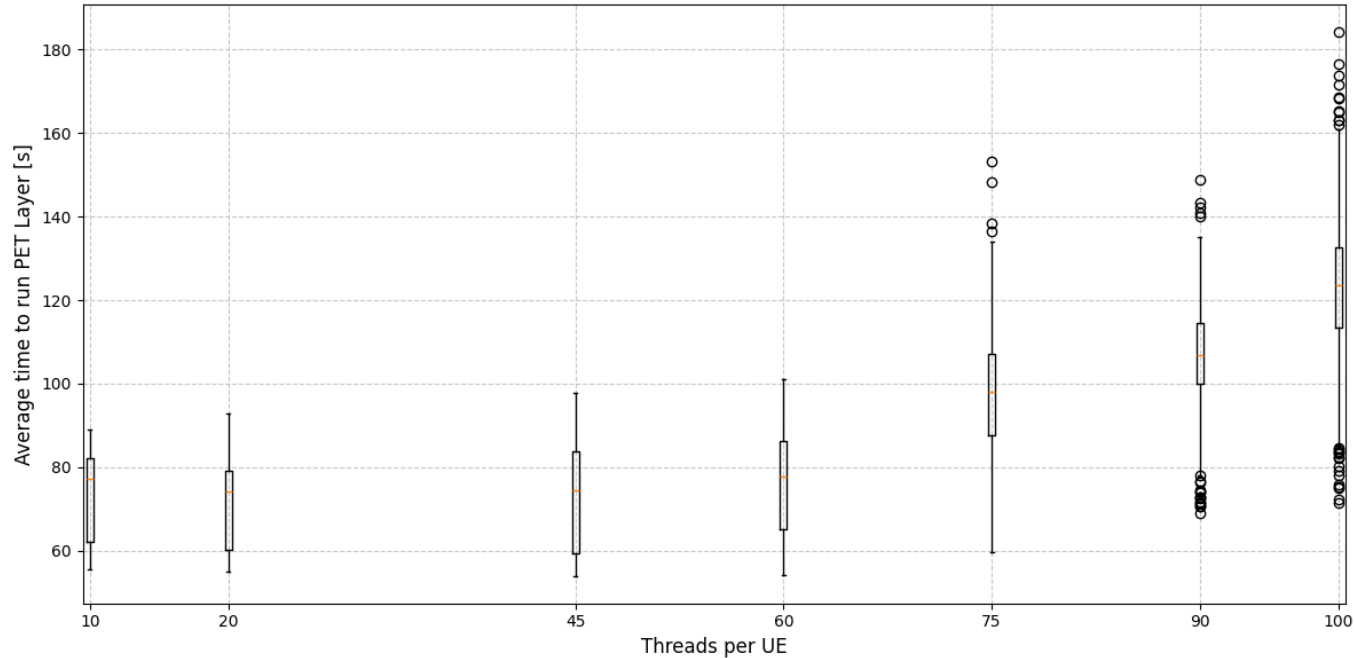
How it was measured

- Measured on UC2: Cloud processing of data for correlation
- Used Default dataset
- Processing of 1 minute (from 10 minutes recorded)
- Performance was measured by multiple runs with different CPU Threads per UE
- Performance was measured with two batches:
 - First batch of runs was performed with PET layer
 - Second batch of tests was performed with GNSS and PET layer

Parameter	Value
Signal duration	10 minutes
Type of signal	synthetic
Sampling frequency	12.5 Msps
Quantization	8
Number of UEs	2
Number of SP	1
Constellations	GPS and Galileo
Signals	L1C/A and E1B/C
SV Tracked	14

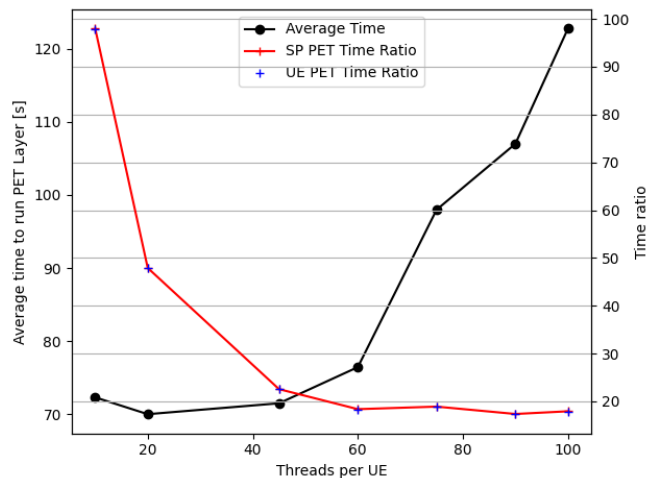
Computational Performance - PET only

Batch one: time to process 1s of signal on a single thread vs number of threads



Computational Performance - PET only

Time to process 1s of signal on a single thread and time processing ratio vs number of threads



Threads	Resident Memory [kB]	Elapsed time (hh:mm:ss)	Time ratio	Average time to run PET Layer [s]
10	1341540	01:37:54	97.90	72.31
20	1139720	47:51.5	47.85	70.02
45	1214744	22:30.6	22.50	75.51
60	1246272	18:19.4	18.32	76.45
75	1537508	18:51.6	18.85	97.98
90	1311200	17:19.3	17.32	106.99
100	1339080	17:53.1	17.88	122.73

Computational Performance - PNT and PET layers

↓ SP

↓ UE

PET



Threads	Resident Memory [kB]	Elapsed time (mm:ss)	Time ratio
90	510716	18:43.4	18,72
92	532776	17:24.3	17,41
94	497904	17:52.7	17,88
95	522220	17:14.0	17,23

Threads	Resident Memory [kB]	Elapsed time (mm:ss)	Time ratio
90	1391028	18:36.55	18,60
92	1363288	17:18.66	17,31
94	1449124	17:47.91	17,79
95	1321600	17:07.53	17,12

PNT



Threads	Resident Memory [kB]	Elapsed time (mm:ss)	Time ratio
90	1005776	12:59.5	12,99
92	1059660	12:38.0	12,63
94	985588	12:09.2	12,15
95	1004032	12:11.2	12,18

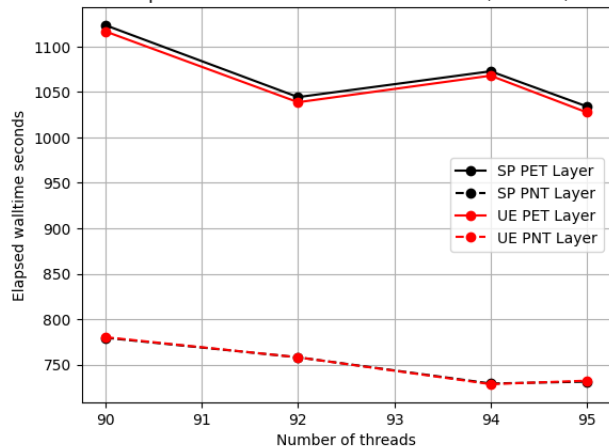
Threads	Resident Memory [kB]	Elapsed time (mm:ss)	Time ratio
90	995796	13:00.1	13,00
92	987412	12:38.0	12,63
94	982192	12:08.5	12,14
95	1006164	12:12.3	12,20

29,41

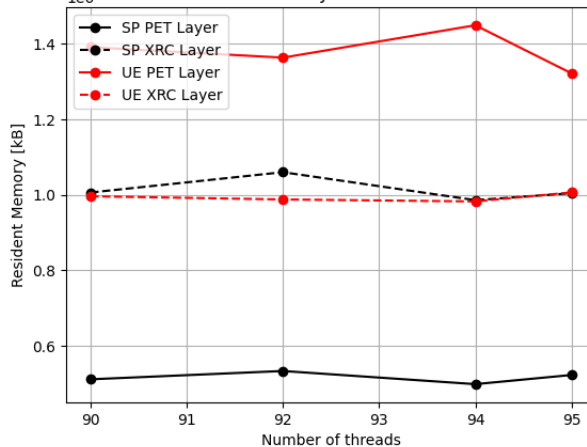
29,32

Computational Performance PNT and PET layers

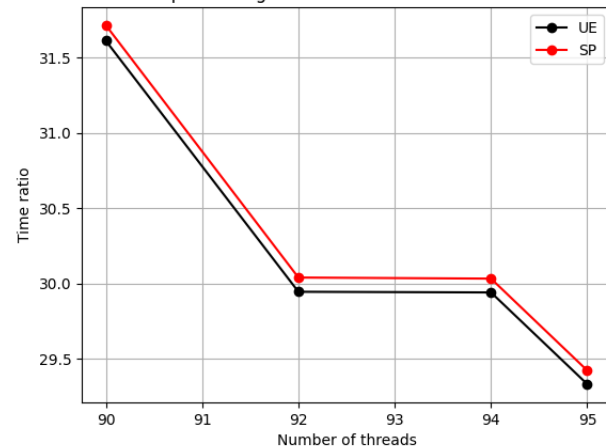
Elapsed wall time vs. number of threads (detailed)



Max Resident memory vs. number of threads

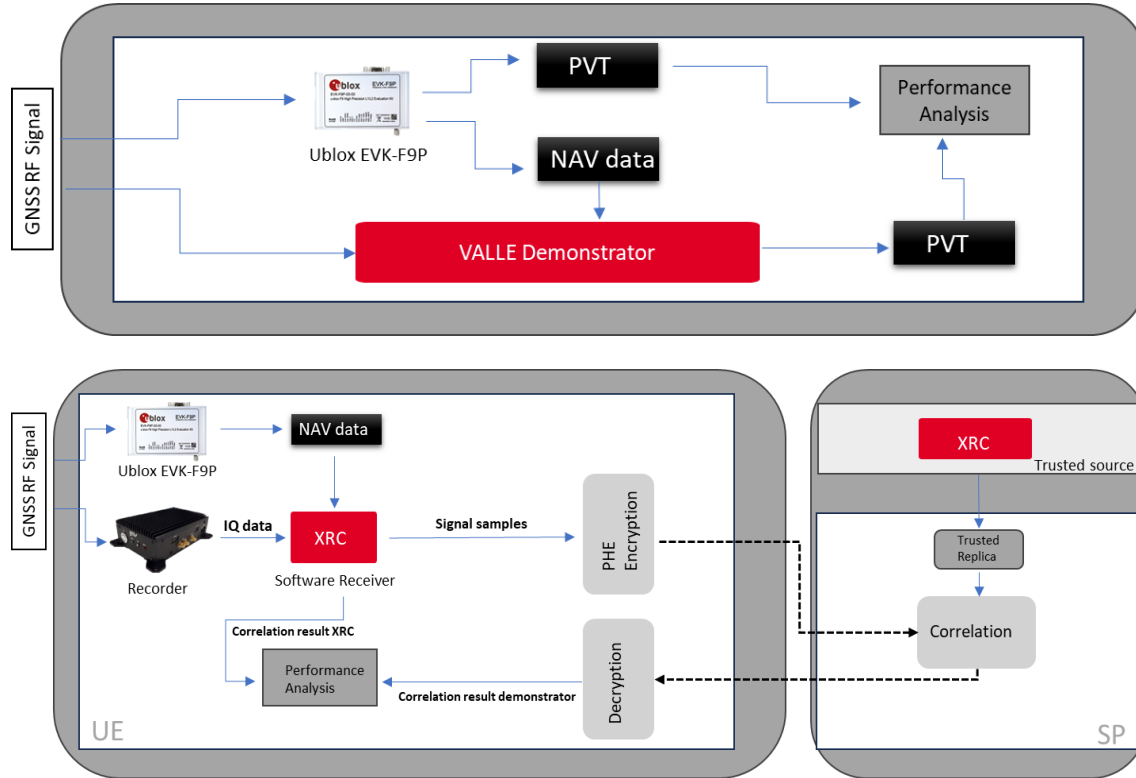


E2E processing time ratio vs. number of threads



Navigation Performance Analysis

PVT and Correlation performance analysis workflow

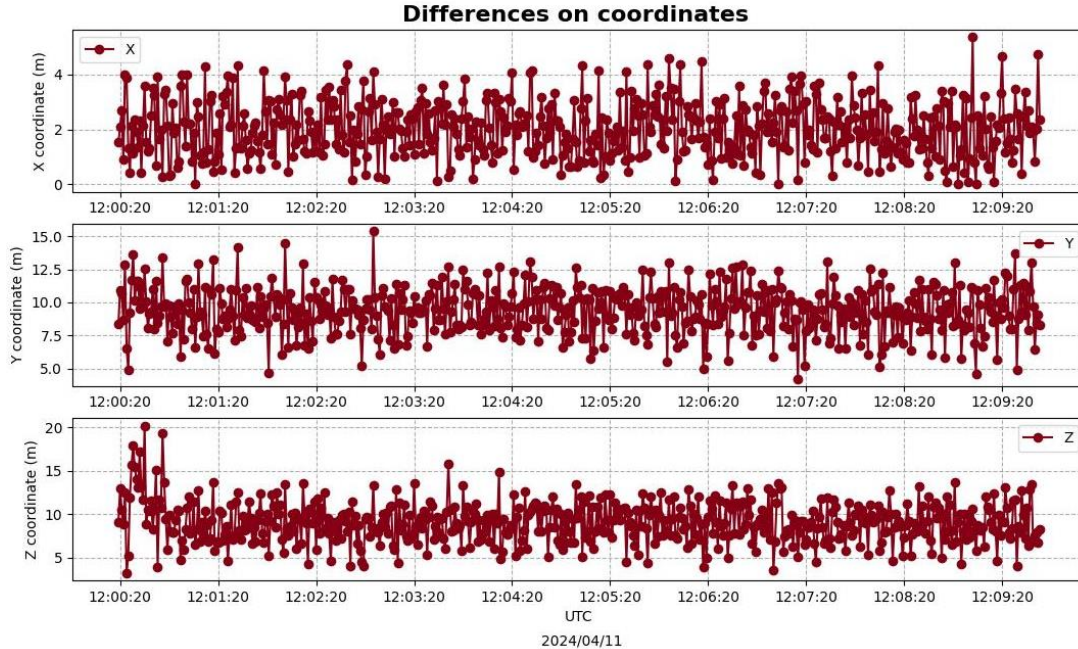


Performance analysis datasets

Dataset	Source	Duration	Signal	Users	Config
DS-1	Simulator	300 s	GPS L1C/A, Galileo E1BC	5	12.5 Msps, 8 bits, 1 channel
DS-2	Simulator	600 s		5	
DS-3	SiS	300 s		2	

Navigation Performance Analysis

Crowd Management Applications Navigation Performance Analysis



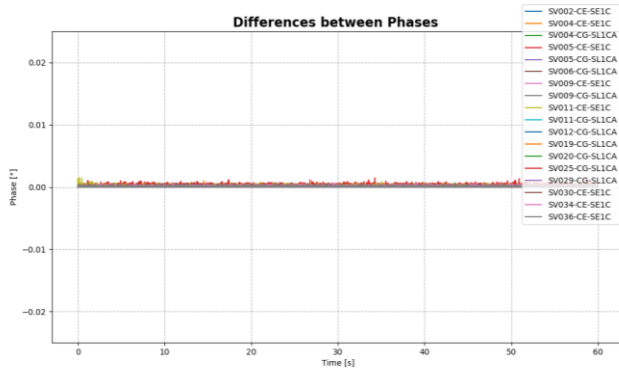
**Coordinates differences for UE5
(Dataset 2)**

	Pos. error (m)	X error (m)	Y error (m)	Z error (m)
Mean	13.664	1.843	9.439	9.558
Min	5.974	0.002	3.980	2.979
Max	22.639	5.208	15.105	17.763
STD	2.638	0.998	1.772	2.318

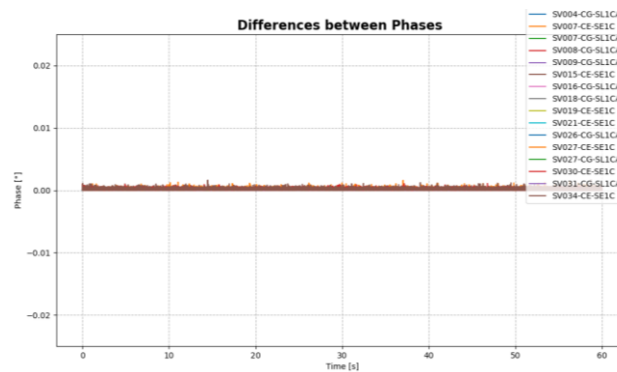
Statistics of all 5 users

Navigation Performance Analysis

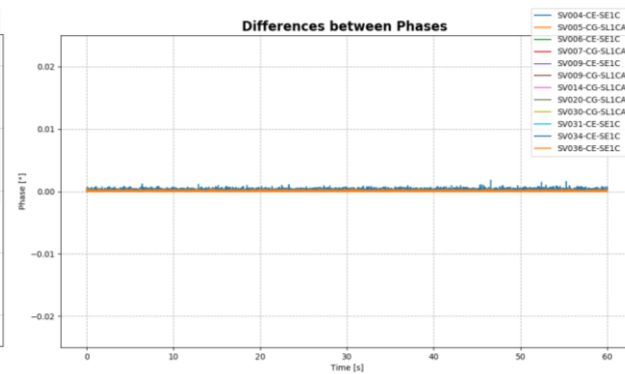
Cloud Processing of Data for Correlation Navigation Performance Analysis



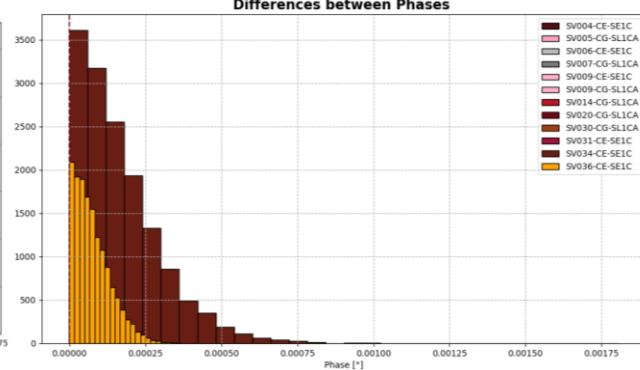
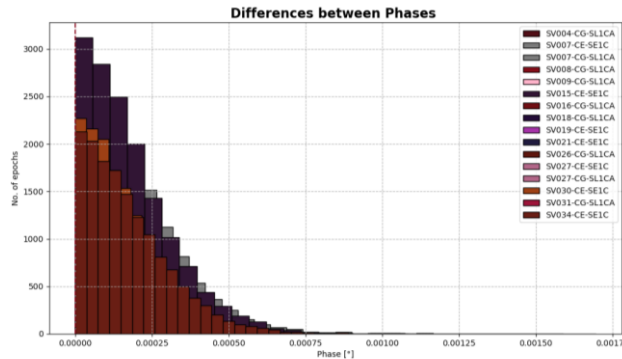
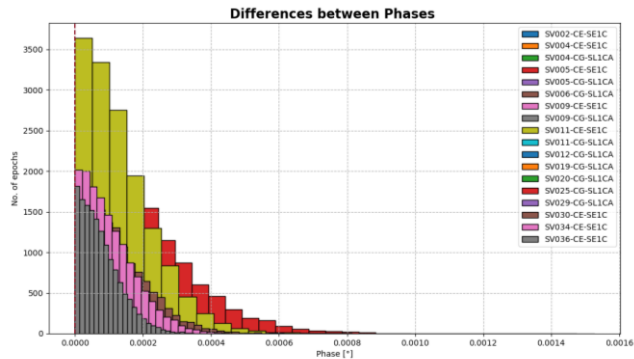
Dataset 1



Dataset 2

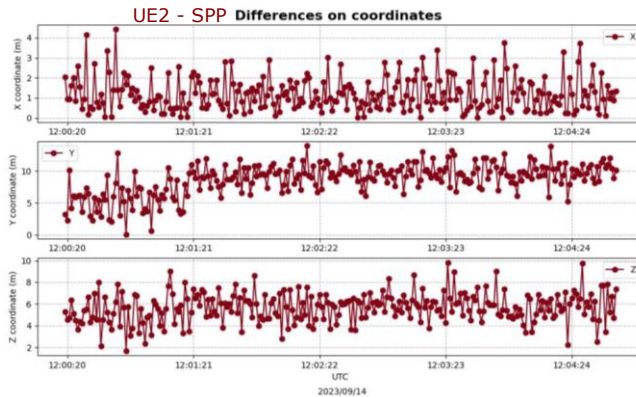
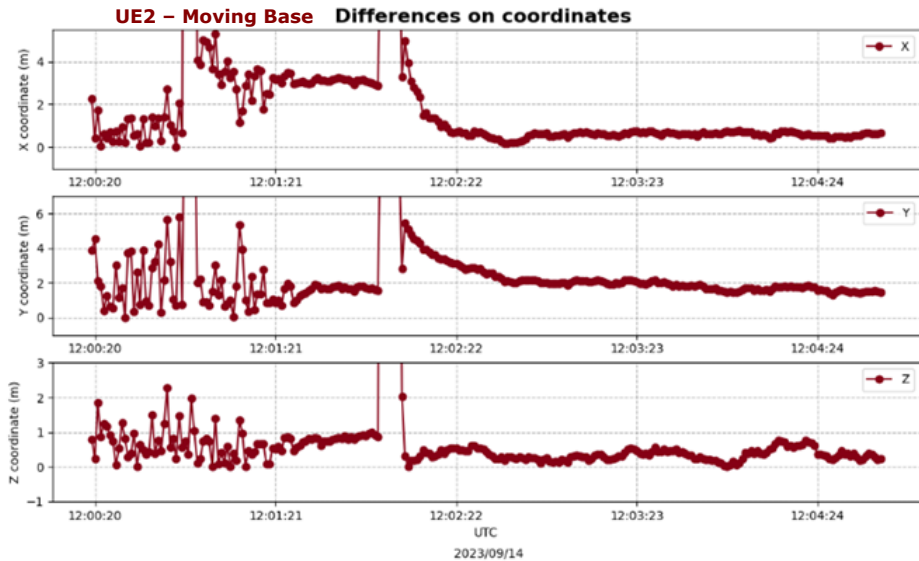
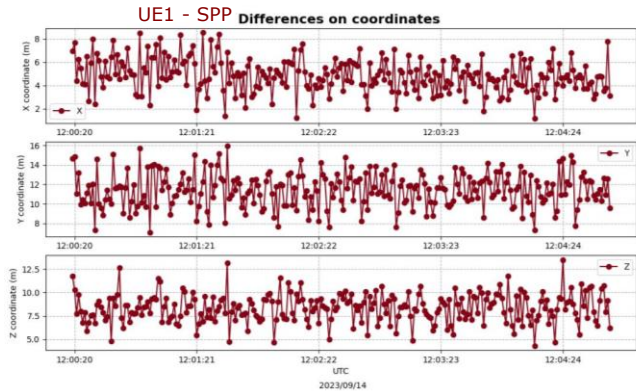


Dataset 3



Navigation Performance Analysis

Secure Collaborative Positioning Navigation Performance Analysis

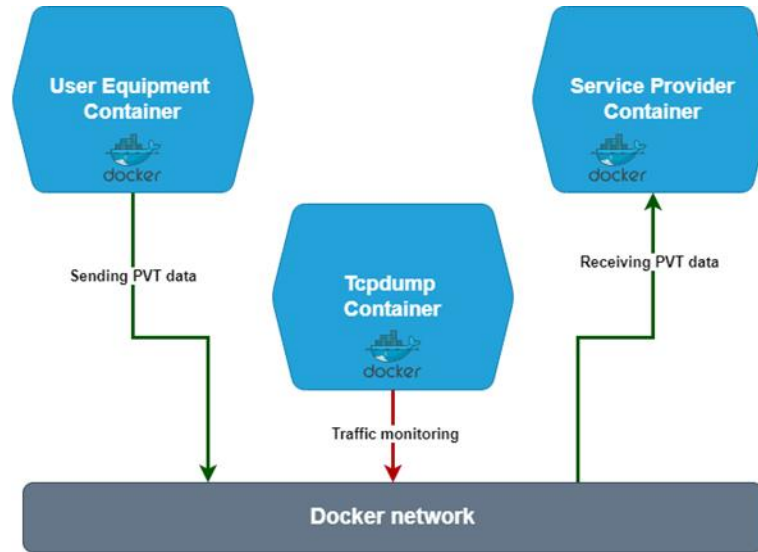


	Pos. error (m)	X error (m)	Y error (m)	Z error (m)
Mean	3.793	2.161	2.457	0.896
Min	0.622	0.016	0.016	0.005
Max	41.574	37.762	17.793	16.156
STD	5.596	4.735	2.389	2.463

UE2 statistics (Moving Base)

Red team analysis

Methodology



Red team analysis

Outcomes and Conclusion

Protocols types: The captured traffic contains only three protocol types: ICMPv6, mDNS, and ARP. These are commonly used for address and hostname resolution at different layers of the network stack.

Traffic patterns: The captured network traffic looks similarly in all the three scenarios

Outcomes: The timeline of the captured packets displays noticeable gaps between them. During the analyzed timeframe the hacking container was in the same docker network with the target containers. The red team analysis analyzed the captured traffic and network protocols used by the Docker containers.

No vulnerabilities were identified in the captured traffic.

Conclusions and Future Work

Project Conclusions

Project Achievements

- ✓ Identified, defined and consolidated a set of use cases for privacy-preserving positioning solutions and/or services based on sharing and processing user PNT data (e.g., intermediate frequency IF or baseband samples, observables, and navigation data, PNT products/trajectory information, etc.).
- ✓ Defined and developed three privacy preserving PNT processing concepts based on the sharing and processing of different types of user PNT data (e.g. IF or baseband samples, observables and navigation data, PNT products/trajectory information, etc.), and assessed the security of the proposed PNT processing concepts.
- ✓ Designed and developed a flexible concept demonstrator
- ✓ Verified and validated the proposed privacy-preserving PNT processing concepts, and benchmarked (processing time and resource usage, latency, robustness) the proposed concepts against standardized solutions

Project Conclusions

Use Cases Achievements

1. The project showed that use of secure multiparty computation for computation of user density is achievable and computationally feasible on a personal computer. The implemented computational algorithm allows for relatively fast securing of users position and sharing the secured data with the service provider for location based services (LBS).
2. The project showed that use of Partially Homomorphic encryption for encryption of IQ samples and performing correlation of IQ samples in the encrypted domain is feasible, computationally achievable on a single server class computer. The conducted study opened the further investigation of improving the existing algorithms to perform the correlation activities in the encrypted space
3. The project showed that use of anonymization techniques over GNSS observables enables a private collaborative positioning solution providing high accuracy PVT solutions.

Future Work

Potential Non-Space Applications

1. Potential non-space applications include implementation of the secure multiparty computation for novel products including crowd management applications, contact tracking, measuring of time specific crowd density in urban and rural areas, IoT and smart city features, etc.
2. Potential non-space applications of correlation of encrypted IQ samples include novel technology allowing secure correlation of GNSS signals without disclosing sensitive user data, measurements etc.
3. Potential non-space applications of anonymization techniques over GNSS observables enable the development of a distributed system performing private collaborative positioning computation based on privately shared observations and providing high accuracy PVT solutions.

The results of VALLE are promising and GMV is interested to pursue the topic of privacy preserving PNT processing concepts.

Work on evolving the Technology Readiness Level is considered and continuation options are being investigated internally in GMV.

Working with ESA

- The development of VALLE was possible thanks to the support provided by ESA and the NAVISP Element 1 programme.
- ESA has provided valuable technical review and guidance at all stages of the project.
- ESA provides access to cutting-edge technology and expertise being at the forefront of GNSS research.
- Privacy Protection for PNT applications and VALLE technology gain increased visibility through the ESA NAVISP mechanisms.

Q & A

Thank you

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