Optimizing Secure Multi-Party Computation in Satellite Proximity Operations



Caroline Fedele University of Florida 7 December 2023







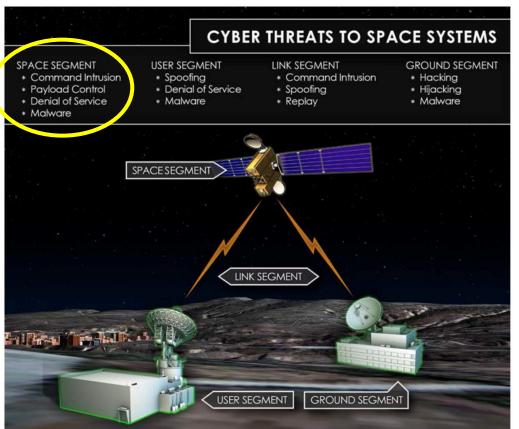






Security in Space





- Needs for both privacy and security in space
- Collisions have occurred

IN-SPACE Cybersecurity

- Growing number of satellites & expanding private sector
- Motivates autonomy needs
- Rendezvous & Proximity
 Operations (RPO)
- Near-field collision avoidance
 and characterization

















General goal: provide evaluation of secure satellite proximity operations using privacy-preserving computation

Demonstrate use of **secure multiparty computation (SMC)**, a method of operating on encrypted data, allowing private **satellite operations** to be conducted between mutually-distrustful agents

Previous work:

- Investigated where data privacy is needed in space
- Implemented SMC into matrix multiplication, attitude optimization, and other algorithms using the Sharemind SMC toolkit (3+ party, secret-sharing based protocol)
- Benchmarked time and memory overhead between each algorithm without and with SMC Current work:
- Implementing SMC into more satellite proximity algorithms, demonstrating improvements
- Benchmarking various overhead measurements of 20+ secure protocol variants using MP-SPDZ













Motivation: RPO

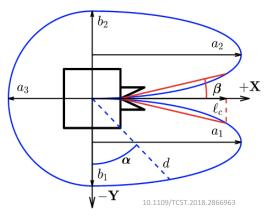


Rendezvous and Proximity Operations (RPO):

- o On-board trajectory operation and replanning
 - E.g. docking, on-orbit servicing/refueling, formation flying
- RPO occurs on-board, autonomously
 - \circ $\$ housed in guidance navigation and control (GNC) unit
- Needed at scales of < 500km between satellites

Ground station vs On-board Control

RPO example: docking



	Ground station	On-Board		
Distance between satellites	1-10 Mm	< 500 km		
Time needed	Days-weeks	< 1 day		
Speed	km /sec	m /sec		
Approach	conjunction analysis	RPO		













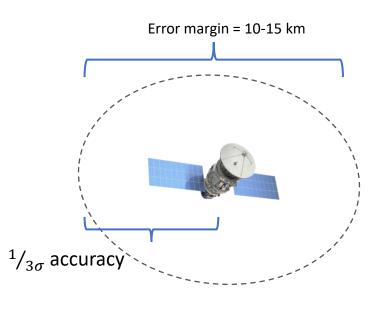
Problem: Capability Inference

Example: Collision Avoidance in RPO

- Minimum data to share with other satellites
 - position, velocity covariance

Stochastic systems

- Probabilistic, not deterministic
- Covariance matrices = quantify uncertainty
 - defined by ellipsoid
- Measure of TRUST, decisions based on accuracy



Problem: knowledge of error margins (covariance matrices) can lead to inferences on satellite capabilities, purpose, etc.

Solution: protect error margins using privacy-preserving computation







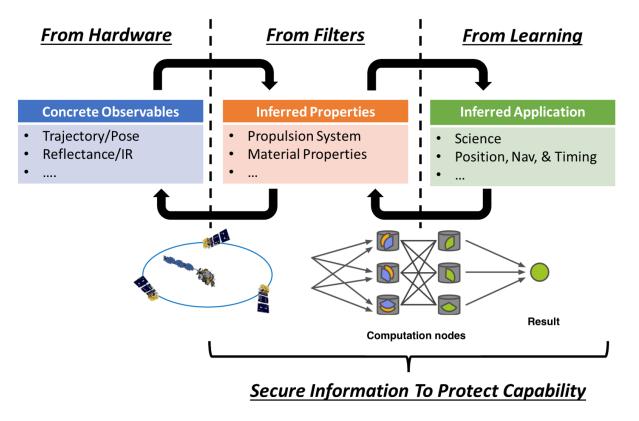








Characterization Problem

















Privacy-Preserving Computation (PPC)

- Allows for data to remain encrypted during computation
- Protects **physical integrity** of satellite during RPO and **data privacy** keeping data encrypted

Secure Multiparty Computation (SMC):

- Cryptographic protocol that allows set of mutually-distrusting parties to jointly compute a function on their inputs, without revealing information about inputs (millionaire's problem)
 - 1. 2-Party Computation (2PC): e.g. Yao's garbled or BMR, binary circuit representation
 - 2. Secret sharing: 3+ parties, arithmetic circuit representation

Homomorphic Encryption (HE):

- Fully or Partially *homomorphic encryption* (FHE/PHE)
- "holy grail" of cryptography, providing strongest privacy guarantees at the cost of efficiency













Mod 2^k

SPDZ2k

N/A

Semi2k

Brain / Rep3

/ PS / SY

Rep3

Rep4

Dealer

Table of supported protocols

Security Models

- Honest vs. dishonest majority assumption of behavior of parties
- Semi-honest vs. malicious corruption passive vs. active adversary

Computation Domain

Mathematical structure of secret info

- Usually ring structure defined by integer operation with modulus or Galois (finite) field
- Binary circuits or arithmetic circuits
 - Mod prime, mod power 2

Underlying Primitives

- Secret Sharing
- Garbled Circuits
- Oblivious Transfer
- Homomorphic Encryption









Security model

majority

majority

majority

majority

Semi-honest.

Malicious, dishonest

Covert. dishonest

dishonest majority

Malicious. honest

Semi-honest, honest

Malicious, honest

Semi-honest, dealer

supermajority



Mod prime / GF(2ⁿ)

MASCOT / LowGear /

CowGear / ChaiGear

Semi / Hemi / Temi /

Shamir / Rep3 / PS / SY

Shamir / ATLAS / Rep3

HighGear

Soho

Rep4

Dealer



	SA	N
Hı.	эΠ	

Bin. SS

Tiny /

Tinier

N/A

SemiBin

Rep3 /

Rep3 /

CCD

Rep4

Dealer

CCD / PS

Garbling

BMR

N/A

BMR

BMR

BMR

N/A

N/A

Yao's GC /



MP-SPDZ vs. Sharemind

Sharemind	MP-SPDZ		
Ease of use for industry & non-security professionals	Prominent tool for academic research uses		
C++ and proprietary SecreC code	Python		
1 SMC approach – linear secret sharing (3+ parties)	Over 30 SMC variants (GC, OT, FHE, SS)		
1 security model (semi-honest)	3 security models (semi-honest, malicious, covert)		
1 trust option (honest majority)	2 trust options (honest or dishonest majority)		
Black box – cannot see or modify source code	White box – can see and modify source code		







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SMC on Satellites

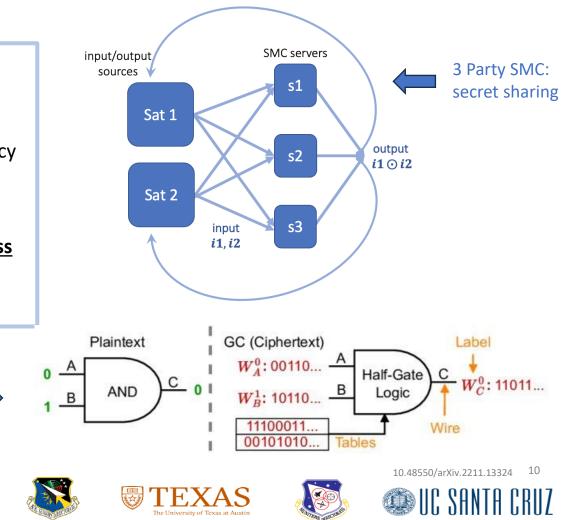
Our purpose

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- Optimize protocol/variants for specific operations, informed by satellite algorithms that need privacy
- Demonstrate reasonable <u>efficiency</u> for each satellite operation
- Guarantees of <u>privacy</u> & <u>correctness</u> for each

2 party SMC:

garbled circuits





Methodology: Software

Security	Parties	Modulo	Protocol
	3	2 ⁶⁴ 128-bit prime 128-bit prime	[AFL ⁺ 16] [AFL ⁺ 16] [CDM00]
Semi-honest	2	2 ⁶⁴ 128-bit prime 128-bit prime 128-bit prime	[DEF ⁺ 19]- [KOS16]- [KPR18]- (semi HE) [KPR18]- (somewhat HE)
Covert	2	128-bit prime	[KPR18]* (semi HE) [KPR18]* (somewhat HE)
Malicious	4	2 ⁶⁴	[DEK20]
Malicious	3	2 ⁶⁴ 2 ⁶⁴ 128-bit prime 128-bit prime 128-bit prime 128-bit prime	[EKO ⁺ 20] (post-sacrifice) [ADEN19] [LN17] (replicated) [CGH ⁺ 18] (replicated) [LN17] (Shamir) [CGH ⁺ 18] (Shamir)
	2	2 ⁶⁴ 128-bit prime	[DEF ⁺ 19] [KOS16]

Integrating SMC into satellite operations

- Testing different RPO algorithms
 - Quadratic Program
 - Conjunction Analysis

Software toolkit

- MP-SPDZ
 - Platform for 30+ SMC operations
 - System of libraries based in python, designed for easy, even comparison between protocols variants

https://eprint.iacr.org/2020/521.pdf











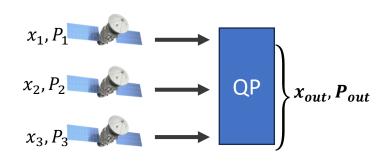


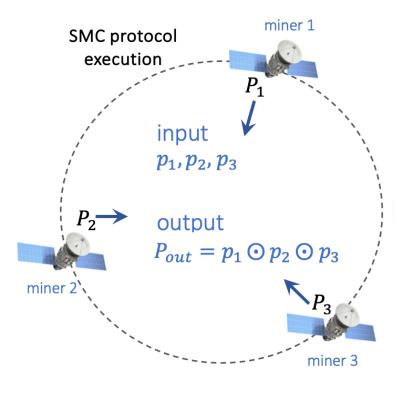
Algorithm 1: Quadratic Program



Quadratic Program: multi-point inspection

- Sensor Fusion optimization algorithm
- Need 3+ parties for 3 dimensional accuracy (secret sharing or homomorphic encryption)
- INPUT: position vector, *x*, and uncertainty matrix, *P*, for each satellite (only P is private)
- OUTPUT: optimized/most accurate {x, P} pair









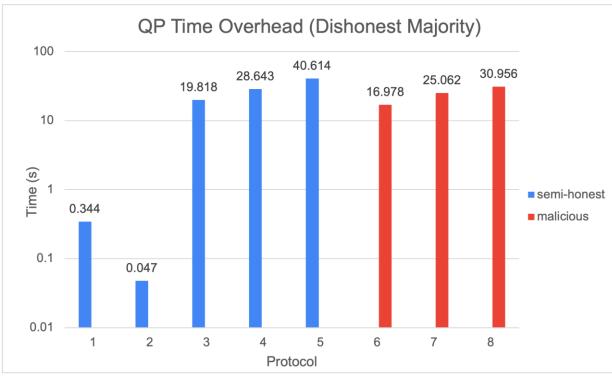












1-8 correspond to different arithmetic circuit protocols

Most efficient \rightarrow #2 = semi2k, modulo 2^k oblivious transferbased protocol





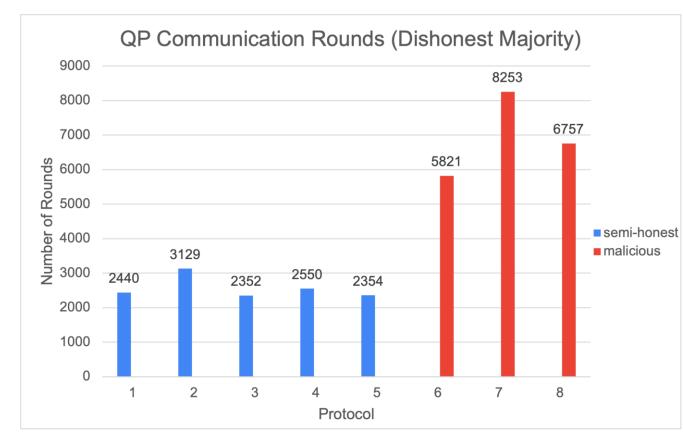












- Higher number of rounds for Malicious model

- significant factor in space applications





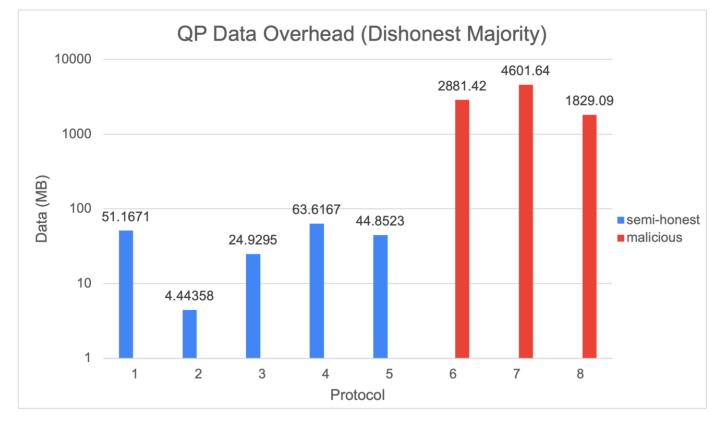












Most efficient time and memory \rightarrow #2 = semi2k, modulo 2^k oblivious transfer-based protocol

















Further motivation for MP-SPDZ

- approx. 1 order of magnitude improvement over Sharemind
- about 2 orders of magnitude greater than without SMC (state-of-the-art)
- QP: need < 10 s to compute. This is < 0.1 s so well within efficiency















Algorithm 2: Alfano's Method



Alfano's Method: conjunction analysis

- Calculate collision probability between two spherical objects
- Assume linear orbital dynamics: one satellite stationary relative to other
- 2 party SMC problem, no trusted 3rd party (Garbled Circuits or Oblivious Transfer)
- INPUT = {x_{1,2}, v_{1,2}, R_{1,2}, C_{1,2} } for satellite 1 and 2, only covariance matrices, C_{1,2}, are private
- OUTPUT = *p*, probability of collision

$$p \leftarrow \frac{1}{2\pi\sigma_x\sigma_y} \int_{-R}^{R} \int_{-\sqrt{R^2 - x^2}}^{\sqrt{R^2 - x^2}} f(x, y) dy dx$$



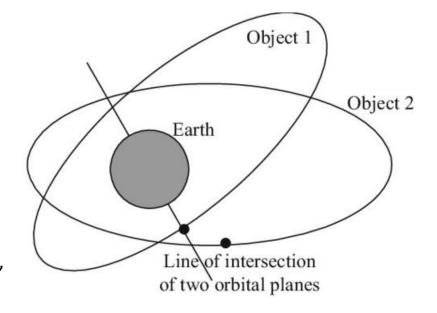






where $f(x,y) = \exp\left[\frac{-1}{2}\left[\left(\frac{x-x_m}{\sigma_x}\right)^2 + \left(\frac{y-y_m}{\sigma_y}\right)^2\right]\right]$





Takeaway





Source: verdict.co.uk

Current Work:

- Optimizing MP-SPDZ protocols for QP algorithm
- Testing Alfano's method with MP-SPDZ
- AIAA (SciTech) Conference paper accepted

Future Work:

- Cybersecurity Conference paper in February (USENIX '24)
- Further examinations of space characterization issue and areas where privacy can be beneficial

















Kevin Butler Tyler Lovelly Chris Petersen Carson Stillman









uke







Evaluation: Matrix Multiplication



100000 $y = 4E - 11x^{4.5234}$ 10000 $y = 4E - 11x^{4.0693}$ 1000 time (log s) 100 10 1 500 1000 1500 2000 2500 0.1 0.01 matrix size (bytes)

Matrix Multiplication

non smc smc

SMC increases time to perform algorithm on each matrix by 1-1.5 orders of magnitude







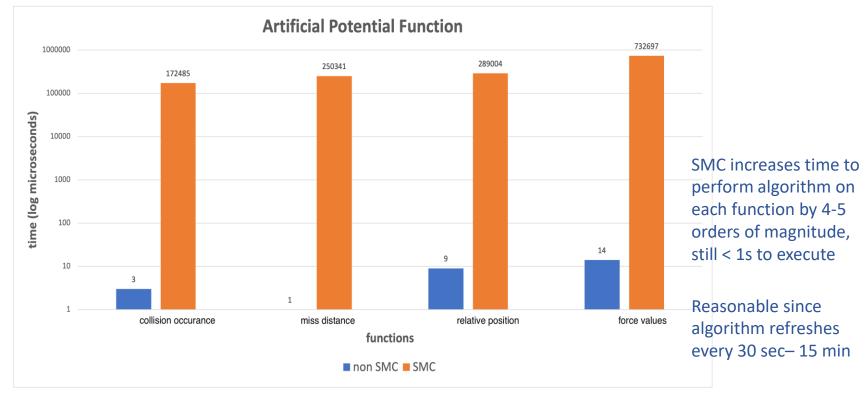






Evaluation: APF





UF FLORIDA





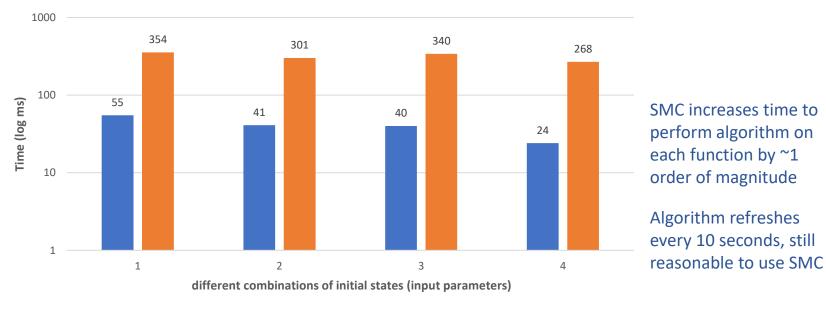




DUC SANTA CRUZ



Evaluation: Optimization



Attitude optimization

non SMC SMC









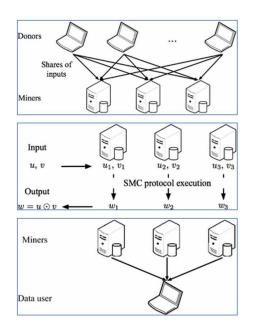


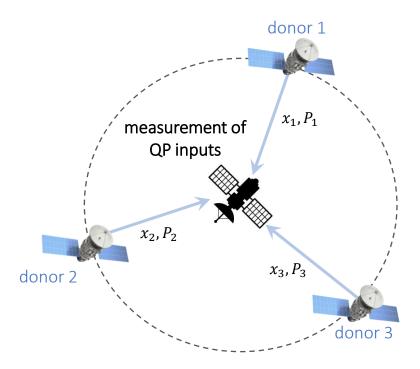




Optimization Algorithm

satellite setup











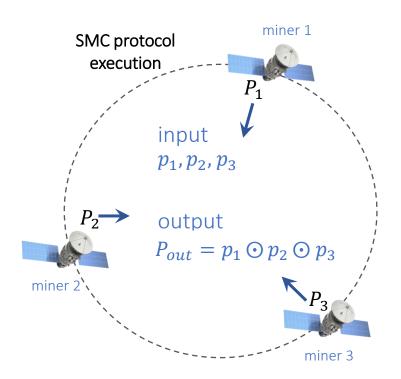
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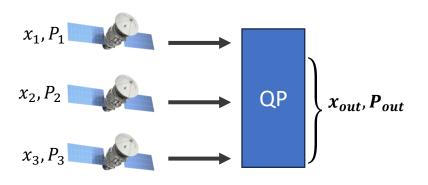




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 (p_1, p_2) (p_1, p_3) (p_2, p_3)







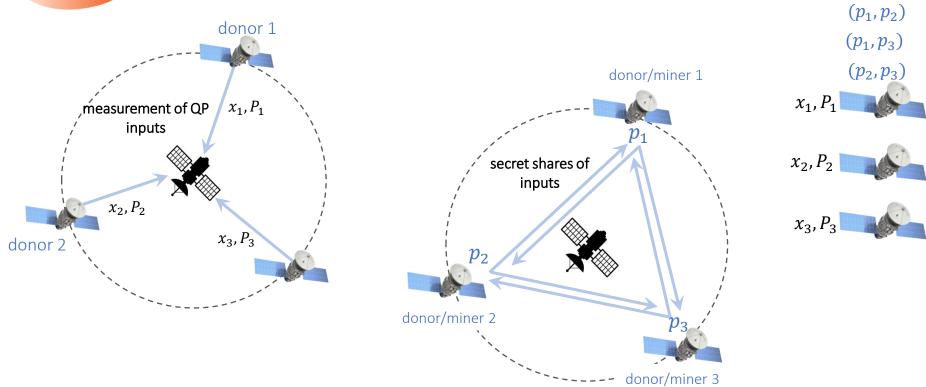








8- ...















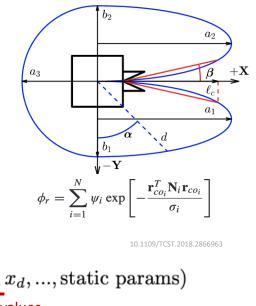
Docking Algorithm



Another example: Artificial Potential Function (APF)

- Scenario: docking & collision avoidance at close range
 - On-board trajectory control
- Linear (relative) equations of motion

Keep-out zone potential



subsystem inputs (vehicle dynamics) f_{x}, f_{y}, τ GNC $\phi_{r} = \sum_{i=1}^{N} \psi_{i} \exp \left[-\frac{\mathbf{r}_{co_{i}}^{T} \mathbf{N}_{i} \mathbf{r}_{c}}{\sigma_{i}} \right]$ (control parameters (static table) \mathbf{x}_{d} human control (on-board flexibility) \mathbf{x}_{d} \mathbf{GNC} \mathbf{v}_{ctrl} $\mathbf{v}_{i} = apf(f_{x}, f_{y}, \tau, x_{d}, ..., static params)$ encrypted values















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				Deputy	Object of investigation
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