Smart Contracts for Crowdsourced Spectrum Sensing

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Upload	Select	Evaluate
Task	Workers	Workers





Upload	Select	Evaluate	Pay
Task	Workers	Workers	Workers

Register	Perform
worker	Task



Upload	Select	Evaluate	Pay
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Register	Perform	Evaluate
worker	Task	Requester





- Trust
- Service fees
- Single point of failure

- Dispute resolving (repudiation of payments)
- Prone to attacks or fraud (free-riders)
- Heterogeneity (e.g., uncalibrated, low-precision sensors)



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Distributed ledger solutions

Workers

Distributed Ledgers for Crowdsourcing

- a decentralized crowdsourcing system
- smart contracts (SC) to implement functions of a crowdsourcing platform
- requires a secure environment which is decentralized, unalterable and programmable: provided by Blockchain networks
- transaction fee for miners who confirm the transaction and support blockchain running persistently [Li 2019]

- Lu, Yuan, Qiang Tang, and Guiling Wang. "Zebralancer: Private and anonymous crowdsourcing system atop open blockchain." *IEEE ICDCS* 2018.
- M. Li et al., "CrowdBC: A Blockchain-based Decentralized Framework for Crowdsourcing," in IEEE Transactions on Parallel and Distributed Systems, 2019.

Challenge: SCs storage and processing limited



Spass: Spectrum Sensing as a Service via Smart Contracts

presented at IEEE DySPAN 2018

(under review, IEEE Transactions on Cognitive Communications and Networking)



Amount

Time





















internet that something happens











Mobile Network Operator (MNO) #2
























A Spass contract



- Smart contract usage is not free
 - Write operations are expensive! (internal storage and manipulation of the contract)
 - Higher cost with increasing complexity of computation

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Operation	Gas	Description
ADD/SUB	3	Arithmetic operation
MUL/DIV	5	Arithmetic operation
ADDMOD/MULMOD	8	Arithmetic operation
AND/OR/XOR	3	Bitwise logic operation
LT/GT/SLT/SGT/EQ	3	Comparison operation
POP	2	Stack operation
PUSH/DUP/SWAP	3	Stack operation
MLOAD/MSTORE	3	Memory operation
JUMP	8	Unconditional jump
JUMPI	10	Conditional jump
SLOAD	200	Storage operation
SSTORE	5,000/20,000	Storage operation
BALANCE	400	Get balance of an account
CREATE	32,000	Create a new account using CREATE
CALL	25,000	Create a new account using CALL

Gas costs: https://docs.google.com/spreadsheets/d/1m89CVujrQe5LAFJ8-YAUCcNK950dUzMQPMJBxRtGCqs/edit#gid=0

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Our proposal: Smart Contracts for Spectrum Sensing (Spass)

(i) what is the cost of running smart contract based spectrum discovery?

(ii) given that both the helpers and the miners have to be paid, under which conditions an MNO can sustain a *profitable business* via smart-contract based spectrum discovery?

(iii) How can the smart-contract catch *free-riders* among the sensing participants?

Spass

- Design goals
- Contract functionality
- Optimal contract parameters
- Malicious helper identification
- Performance
- Business feasibility of Spass



















• Contract duration: a certain time period







Backhaul

- Contract duration: a certain time period
- SUN uploads the contract to Ethereum and receives a unique ETH address
- SUN broadcasts the contract address over the air



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 - accuracy in PU detection, false alarms
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Message flow

- Spectrum sensing
 - Energy detection (hard decision: 0 or 1)
 - PU activity report (binary array)
 - SUN sensing decision fusion: majority voting
- Helper accuracy verification
 - Compressed reports
 - Malicious helper detection and payment
 - Goal: High malicious helper detection accuracy, low # blacklisted honest helpers
 - Payment to only whitelisted-helpers



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How to *design* a smart contract so that SUN maximises its profit?



Spass

Assumptions

- Primary User (PU)
 - active with probability p₁ and inactive with p₀
 - known by SUN and helpers
- Homogenous helpers with identical sensing cost and accuracy
 - Probability of detection (P_d)
 - Probability of false alarm (P_f)
- Malicious helpers with identical sensing accuracy: free riders (no energy consumption for sensing or colluding)
 - do not sense the spectrum but generate data using their statistical knowledge of po
 - Fraction of malicious helper population known by the SUN
 - Malicious helpers do not collude and do not change their strategy

SUN profit

- **Income**: Monetary gain accumulated over all verification rounds by serving its customers over the opportunistic spectrum
- **Payment**: Helpers sensing service and the smart contract cost









Assumption: write operation is the dominant cost in Ethereum usage





We find the operation range of an SU network in which it can make profit from Spass: net profit>0 $\Delta\Upsilon=\Upsilon^+-\Upsilon^->0$

When is Spass profitable?



$$\kappa > \frac{R_s(\mu_{eth}/\beta + \mu_s) \times H}{\mu \times \mathcal{U}^{\text{Spass}} \times B}$$



Impact of compression β under H = 4.

$$\mathbf{P1:}\max_{V,H} \left(V \sum_{v=1}^{N_V} \mu \kappa \mathcal{U}_v - (H - \tilde{M}_v) R_s \left(\frac{\mu_{eth}}{\beta} + \mu_s\right) \right) \tag{8}$$

$$N_V = T_c/V$$

$$\mathcal{U}_v = p_0(1 - \bar{p_f}(H))$$
(10)

$$\bar{p}_f(H) = \sum_{j=0}^{H} {H \choose j} \psi^j (1-\psi)^{H-j} p_f(H,j)$$
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$$p_{f}(H,j) = \sum_{K-\lceil H/2 \rceil}^{H} \sum_{i=0}^{\min(H,j)} {j \choose i} (p_{f}^{m})^{i} (1-p_{f}^{m})^{j-i} \quad (12)$$
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$$S = \frac{R_s V}{\beta} \tag{16}$$

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Utility: probability of detecting the spectrum opportunity

$$P1: \max_{V,H} \left(V \sum_{v=1}^{N_{V}} \mu \kappa \mathcal{U}_{v} - (H - \tilde{M}_{v}) R_{s} (\frac{\mu_{eth}}{\beta} + \mu_{s}) \right)$$
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Expected utility if H helpers sense

Expected false alarm probability

False alarm probability under j malicious helpers and total H helpers

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- Utility: probability of detecting the spectrum opportunity
- Regulatory requirements on sensing accuracy are satisfied

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Fraction of malicious helpers in all population q_d , q_f : true and false detections

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R_{s :} Rate of sensing (bps) ß: compression factor

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 - Sensing report size under compression factor beta

But, we do not have a closed formula for malicious helper detection accuracy!

Contracts published









- Requirement: After the first phase, all malicious helpers are (must be) detected by our MHD algorithm
- Malicious helpers do not change their strategy

- Now, decision on H₀, H₁, T₀
 - T_{0:} the minimum duration that the MHD algorithm needs for detection of all malicious helpers
 - We find T₀ via Monte Carlo simulations
 - H_0 , H_1 : exhaustive search O(N²) where N is number of candidates

P2:
$$\max_{T_0, H_0, H_1} T_0(\mu \kappa \mathcal{U}_0 - H_0 R_s(\frac{\mu_{eth}}{\beta} + \mu_s)) + (T_c - T_0)(\mu \kappa \mathcal{U}_1 - H_1 R_s(\frac{\mu_{eth}}{\beta} + \mu_s))$$
(18)

$$T_0 \leqslant T_c \tag{19}$$

$$\mathcal{U}_0 = p_0(1 - \bar{p_f}(H_0)) \tag{20}$$

$$\mathcal{U}_1 = p_0(1 - p_f(H_1)) \tag{21}$$

$$H_0 \geqslant H_1 \tag{22}$$

$$p_f(H_1) \ge p_f^* \text{ and } p_d(H_1) \ge p_d^*$$

$$\bar{p_f}(H_0) \ge p_f^* \text{ and } \bar{p_d}(H_0) \ge p_d^*$$
(23)
(24)

$$S_{\min} \geqslant \frac{R_s T_0}{\beta}$$

(25)

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Might affect malicious helper detection algorithm's accuracy
How to decrease cost of Spass?



How to decrease cost of Spass?



How to decrease cost of Spass?



• Higher compression when low/high PU idle

Spass

- Design goals
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Business feasibility of Spass



Spass

- Design goals
- Contract functionality
- Optimal contract parameters
- Malicious helper identification
- Performance
- Business feasibility of Spass



Honest helper model: Identical sensing accuracy P^hd, P^hfa

Malicious helper model: Knows P₀, Generates fake sensing bits without performing sensing 1 with prob $\alpha_1 = (1 - p_0)$. $P^m_f = p_0 \alpha_1 = p_0 (1 - p_0)$ $P^m_d = (1 - p_0)^2$.



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Distance between two helpers

normalized Hamming distance of two helper reports



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- Distance between two helpers
 - normalized Hamming distance of two helper reports
- Helper score: x-percentile of its distance from other helpers



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- Distance between two helpers
 - normalized Hamming distance of two helper reports
- Helper score: x-percentile of its distance from other helpers

If 50-percentile: Score-A = [0.2, 0.3, 0.5]

Key insight of CHI

- Honest helpers with similar sensing reports
 - Short distance from each other
 - Similar and low scores
- Malicious helpers (do not collude and do not sense the spectrum)
 - High distance from each other and from honest helpers
- Two clusters
 - Honest helper cluster and malicious
 helper cluster



(b) Expected distance between two helpers according to their type with increasing p_0 .

0

Helper scores

1

K-means clustering for K=1 0
1

Helper scores



Helper scores









• Is the clustering accuracy better above a threshold for K=2 compared to K=1?



• Is the clustering accuracy **better above a threshold** for *K*=2 compared to *K*=1?



• Is the clustering accuracy better above a threshold for *K*=2 compared to *K*=1?







High inertia difference indicates the existence of malicious helper(s)



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Performance

Business feasibility of Spass

- Optimal number of helpers
- Robustness of CHI to malicious helpers
- Impact of lossy compression
- Sensing accuracy

Spass

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Performance

Business feasibility of Spass

Optimal number of helpers



Fig. 8. Impact of increasing p_0 on optimal number of helpers for various ψ values. Following parameters are used: $R_s = 5, \mu_{eth} = 0.1, \mu_s = 0.05, \mu = 1$, and $\beta = 10, \kappa = 10$.

- Higher P₀, higher number of helpers
- More malicious helpers expected, more helpers need to be selected



CHI achieves high malicious helper detection accuracy



CHI achieves high malicious helper detection accuracy





Almost always no false alarms if report size is bigger than 100 bits



Minimum report size needed



Spectrum discovery with very high accuracy



In a nutshell

- Spass: https://github.com/zubow/Spass_contract
 - Spectrum sensing as a service using smart contracts (Ethereum)
 - Cost of smart contract usage and helpers sensing service
 - Spass becomes profitable under some conditions
 - Optimal number of helpers and verification round durations
- Possible directions:
 - More sophisticated malicious users
 - Pricing and reward assignment based on helper capabilities
 - Computation cost should also be considered
Crowdsourcing for spectrum sensing

- Spectrum monitoring for policy making and misuse detection
 - SpecGuard
- Radio Environment Map construction (spectrum sensing provider)
 - SpecSense
- Pricing for crowdsensing
 - Han et al.
- Malicious sensor identification
 - Catch Me If You Can
- Privacy aspects
 - DPSense

- Jin, Xiaocong, et al. "Specguard: Spectrum misuse detection in dynamic spectrum access systems." *IEEE TMC* 2018
- Chakraborty, Ayon, et al. "Specsense: Crowdsensing for efficient querying of spectrum occupancy." *IEEE INFOCOM 2017*
- DPSense: Differentially Private Crowdsourced Spectrum Sensing, ACM CCS 2018
- Ying, Xuhang, Sumit Roy, and Radha Poovendran. "Pricing mechanisms for crowd-sensed spatial-statistics-based radio mapping." *IEEE Transactions on Cognitive Communications and Networking*, 2017.
- Han, Kai, He Huang, and Jun Luo. "Quality-Aware Pricing for Mobile Crowdsensing." *IEEE/ACM TON* 2018
- Li, Husheng, and Zhu Han. "Catch me if you can: An abnormality detection approach for collaborative spectrum sensing in cognitive radio networks." *IEEE Transactions on Wireless Communications* 2010. 55

The future is **unlicensed**, **diverse**, and **decentralized**



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Talk Abstract

To cope with the huge increase in cellular data traffic, mobile network operators (MNO) consider using also other spectrum bands which are under-utilized spatiotemporally. To discover such spectrum opportunities, several studies suggest exploiting the ubiquity of mobile devices to perform spectrum sensing. However, current solutions for crowdsourced spectrum sensing overlook the fact that mobile devices lack the incentives to sense without certain benefits as spectrum sensing results in energy and CPU overhead. To encourage participation, we explore the use of smart contracts running on a distributed ledger (e.g., Ethereum) for an MNO to announce its need for spectrum sensing, the interested sensors to register themselves as candidate sensors, and MNO to pay the selected sensors for their service. Despite the simplicity of the idea, the realization requires more thoughts as smart-contracts are limited in their storage and processing capacity, and incurs a high cost for the write operation. I will introduce our proposal Spass: spectrum sensing as a service via smart contracts and address the following questions: (i) what is the cost of running smart contract based spectrum discovery? and (ii) given that both the helpers and the miners have to be paid, under which conditions an MNO can sustain a profitable business via smart-contract based spectrum discovery? (iii) How can the smart-contract catch free-riders among the sensing participants?

Bio:Suzan Bayhan is a senior researcher at TU Berlin and a docent in Computer Science at the University of Helsinki. She got her Ph.D. from Bogazici University in 2012 and worked as a post-doc at the University of Helsinki between 2012-2016. She was on N2Women board between 2017-2018. Her current research interests are resource allocation for wireless networks, spectrum sharing, and edge/fog/cloud computing. Suzan will join the University of Twente on September 2019 as an Assistant Professor.