

Multi-Leader Multi-follower Stackelberg Game in Mobile Blockchain Mining

# WWW '20

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- **Q2** MODEL DESIGN
  - 03 EXPERIMENT
  - © 04 CONCLUSION



# 问题背景

Paper Background



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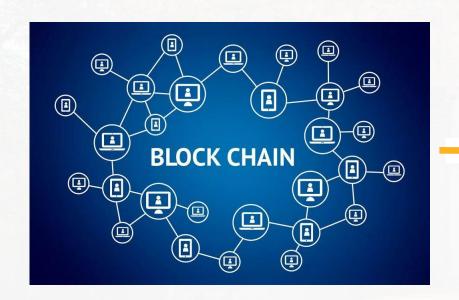
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# 背景介绍



#### ■ Block Chain





区块链技术与比特币



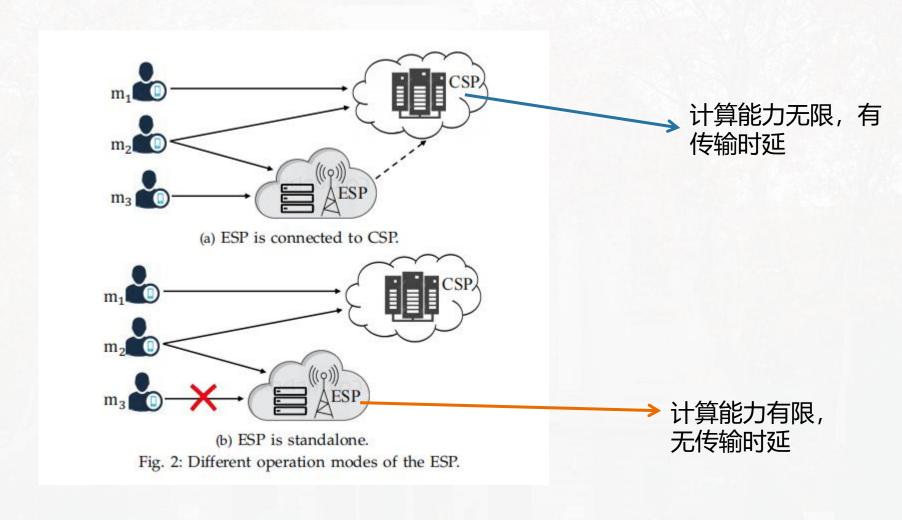












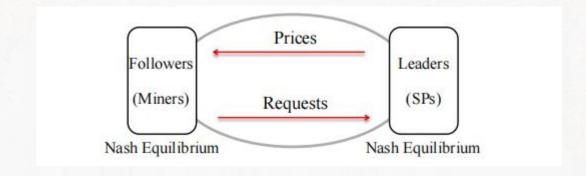
两种服务模型



# **Stackelberg Equilibrium**



(1) 0 + S Ø 0



斯塔克伯格模型

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(1)



## **Motivation and contributions**



- 我们提出了来一个斯塔克伯格模型来解决移动区块链挖掘网络中一个基于价格的计算资源管理问题
- 我们基于其提供的不同质量和价格将ESP和CSP分开,实现资源的最优配置
- 分析了两种边缘操作模式的斯塔克尔伯格平衡(SE)的存在唯一性,在此基础上提出了得到SE解的算法
- 分析了在特殊条件下(矿工同质),SE解的算法
- 我们研究了矿工人数不确定性的影响



模型设计





#### Problem 2 (SP SUBGAME: OPSP).

maximize 
$$V_e = (P_e - C_e) \cdot E$$
 where  $E = \sum_{i=1}^{N} e_i$  (2a)

maximize 
$$V_c = (P_c - C_c) \cdot C$$
 where  $C = \sum_{i=1}^{N} c_i$  (2b)

#### SPs的效益函数

Problem 1 (MINER SUBGAME: OPMINER).

maximize 
$$U_i = R \cdot W_i - (P_e \cdot e_i + P_c \cdot c_i),$$
 (1a)

subject to 
$$P_e \cdot e_i + P_c \cdot c_i \le B_i$$
,  $e_i \ge 0$ ,  $c_i \ge 0$ . (1b)

Miners的问题规划

$$V_e(P_e^*, \mathbf{E}^*) \ge V_e(P_e, \mathbf{E}^*), \forall P_e, \tag{3a}$$

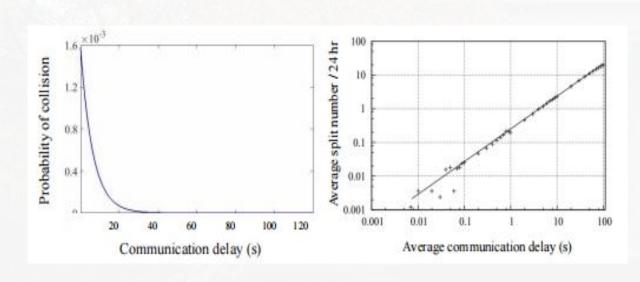
$$V_c(P_c^*, \mathbf{C}^*) \ge V_c(P_c, \mathbf{C}^*), \forall P_c,$$
 (3b)

$$U_i(e_i^*, c_i^*, P_e^*, P_c^*) \ge U_i(e_i, c_i, P_e^*, P_c^*), \forall i.$$
 (3c)

均衡的定义







传输时延对Wi的影响

$$W_i^e(r_i, \mathbf{r}_{-i}) = e_i/S + e_i \sum_{j \neq i} \beta c_j/ES, \qquad (4)$$

$$W_i^c(r_i, \mathbf{r}_{-i}) = c_i/S - c_i \sum_{j \neq i} \beta e_j/ES.$$
 (5)

#### Wi的定义

$$\begin{split} \sum_{i=1}^{N} W_i &= \sum_{i=1}^{N} (W_i^e + W_i^c) \\ &= \sum_{i=1}^{N} \left[ e_i / S + c_i / S \right] \\ &+ \beta \sum_{i=1}^{N} \left[ e_i (C - c_i) / ES + c_i (E - e_i) / ES \right] \\ &= 1 + \beta \sum_{i=1}^{N} (e_i C - c_i E) / ES = 1. \end{split}$$





#### 1.在Connected Mode中,假定计算请求被传输到CSP的概率为h

如果计算请求被ESP接纳的获胜概率(r=[ei,ci]):  $W_i^h = (e_i + c_i)/S + \beta(e_i C - c_i E)/ES$ .

如果计算请求被ESP传输给CSP的获胜概率(r=[0,ei+ci]):  $W_i^{1-h} = (1-\beta)(e_i+c_i)/S$ .

总体而言, 获胜概率为:  $W_i = h \cdot W_i^h + (1-h) \cdot W_i^{1-h} = (1-\beta)(e_i + c_i)/S + \beta h e_i/E$ ,

2.在Standalone Mode中,假定计算请求被接纳的概率为h

如果计算请求被ESP接纳的获胜概率(r=[ei,ci]):  $W_i^h = (e_i + c_i)/S + \beta(e_i C - c_i E)/ES$ .

如果计算请求被ESP传输给CSP的获胜概率(r=[0,ci]), S=S-ei:  $W_i^{1-h} = (1-\beta)c_i/(S-e_i)$ .



## 模型求解-存在性证明



*Proof.* Denote the equivalent variational inequality (VI) problem  $VI(K, F) \equiv NEP(X, U)$ , where

$$F := (\nabla_i U_i)_{i=1}^n, \quad \mathcal{X} = ([e_i, c_i]^{\intercal})_{i=1}^n, \quad \mathcal{U} = (U_i)_{i=1}^n,$$

$$\mathcal{K} := \mathcal{K}_1 \times \mathcal{K}_2 \times \cdots \times \mathcal{K}_n, \qquad (11)$$

$$\mathcal{K}_i := \{(e_i, c_i) | P_e \cdot e_i + P_c \cdot c_i \leq B_i, e_i \geq 0, c_i \geq 0\}.$$

#### 转换为变分不等式问题

$$\begin{split} U^i_{ee} &= -\left(R(1-\beta)/S^2 + \beta h/E^2\right) \cdot \left(R(1-\beta)/S^2 + \beta h/E^2\right) \\ &+ \left(R(1-\beta)/S + 2\beta h/E - P_e\right) \cdot \left(\beta h/E^3 - R(1-\beta)/S^3\right), \\ U^i_{ec} &= R(1-\beta)/S \cdot \left(R(1-\beta)/S^2 + \beta h/E^2\right) \\ &+ 2(R(1-\beta)/S + \beta h/E - P_e)R(1-\beta)/S^3 \\ &- \left(R(1-\beta)/S^2 - 2R(1-\beta)/S^3 \cdot c_i\right), \\ U^i_{ce} &= \left(-R(1-\beta)/S^2 - R(1-\beta)/S^3 \cdot e_i\right) \\ &+ \left(-R(1-\beta)/S^3 + 2R(1-\beta)/S^3 \cdot c_i\right), \\ U^i_{cc} &= 2R(1-\beta)/S^3 \cdot e_i - 2(R(1-\beta)/S^2 - R(1-\beta)/S^3 \cdot c_i), \\ \text{Since } \det(\mathcal{H}) &= U^i_{ee} \cdot U^i_{cc} - U^i_{ec} \cdot U^i_{ce} > 0, \, \forall [e_i, c_i]^\intercal \in \mathcal{K}_i, \end{split}$$

Denote  $\mathcal{H}^i$  for the Hessian matrix of  $U_i$  as below.

$$\mathcal{H} := \begin{bmatrix} U_{ee}^i & U_{ec}^i \\ U_{ce}^i & U_{cc}^i \end{bmatrix}, \tag{18}$$

函数的Hession矩阵



Since  $det(\mathcal{H}) = U_{ee}^i \cdot U_{cc}^i - U_{ec}^i \cdot U_{ce}^i > 0, \forall [e_i, c_i]^\intercal \in \mathcal{K}_i$ 

Hession矩阵正定, 转换是有效的



## 模型求解-唯一性证明



Claim 2: There is at most one NE for Problem 1e. To show the uniqueness of the NE point, we first introduce the matrices  $\mathcal{J}_{low}$ , defined as

$$[\mathcal{J}_{low}]_{ij} := \inf_{x \in \mathcal{K}} \begin{cases} |\nabla_{ii}^2 U_i|, & \text{if } i = j, \\ -\frac{1}{2}(|\nabla_{ij}^2 U_i| + |\nabla_{ji}^2 U_j|), & \text{else.} \end{cases}$$
(21)



$$\begin{split} \nabla_{ii}^{2}U_{i} = & U_{ee}^{i} + U_{cc}^{i} \\ = & R[-8(1-\beta)(S-e_{i}-c_{i})/S^{3}] - 2\beta(E-e_{i})/E^{3}, \\ \nabla_{ij}^{2}U_{i} = & \nabla_{ji}^{2}U_{j} \\ = & R(1-\beta)[1-2(S-e_{i}-c_{i})]/S^{2} + h\beta(2e_{i}-E)/E^{3}. \end{split}$$
 (22a)

# 解是唯一的

$$a_{12} + \sqrt{a_{11}a_{22}} = \inf_{\substack{(e_1, c_1) \in \mathcal{K} \\ (e_1, c_1) \in \mathcal{K}}} R(1 - \beta)[1 - 2(S - e_i - c_i)]/S^2 + h\beta(2e_i - E)/E^3 - 8(1 - \beta)\sqrt{\prod_{i=1,2} (S - e_i - c_i)}/S^3 > 0.$$

$$\mathcal{J}_{low} = \begin{bmatrix} a_{11} & a_{12} \\ a_{12} & a_{22} \end{bmatrix}, \tag{23}$$

$$a_{11} = \inf_{(e_1, c_1) \in \mathcal{K}} |\nabla_{11}^2 U_1|, a_{22} = \inf_{(e_2, c_2) \in \mathcal{K}} |\nabla_{22}^2 U_2|, \quad (24)$$

$$a_{12} = \left(-\frac{1}{2}\right) \inf_{(e_1, c_1) \in \mathcal{K}}^{(e_2, c_2) \in \mathcal{K}} (|\nabla_{12}^2 U_1| + |\nabla_{21}^2 U_2|). \tag{25}$$



## 模型求解



$$L_i = R \cdot [(1 - \beta)(e_i + c_i)/S + \beta h e_i/E] - (P_e \cdot e_i + P_c \cdot c_i)$$
$$- \lambda_1(P_e \cdot e_i + P_c \cdot c_i - B_i) + \lambda_2 e_i + \lambda_3 c_i, \tag{12}$$
 and the complementary slackness conditions are

$$\lambda_1(P_e \cdot e_i + P_c \cdot c_i - B_i) = 0, \lambda_2 e_i = 0, \quad \lambda_3 c_i = 0, \quad \lambda_1 > 0, \lambda_2, \lambda_3, e_i, c_i \ge 0.$$
(13)

#### KKT条件



$$e_{i} = \sqrt{\frac{h\beta E_{-i}R}{(1+\lambda_{1})(P_{e}-P_{c})}} - E_{-i},$$
(14)

$$c_{i} = \sqrt{\frac{R(1-\beta)(E_{-i}+C_{-i})}{(1+\lambda_{1})P_{c}}} - \sqrt{\frac{h\beta E_{-i}R}{(1+\lambda_{1})(P_{e}-P_{c})}} - C_{-i},$$

$$B_i = P_e e_i + P_c c_i$$
, where  $E_{-i} = \sum_{j \neq i} e_j$ ,  $C_{-i} = \sum_{j \neq i} c_j$ .

Solving Eq. (14) yields that

$$1 + \lambda_1 = \left[ \frac{(P_e - P_c)\sigma_1 \sqrt{E_{-i}} + P_c \sigma_2 \sqrt{E_{-i} + C_{-i}}}{B_i + P_c C_{-i} + P_e E_{-i}} \right]^2, \quad (15)$$

where:  $\sigma_1^2 = h\beta R/(P_e - P_c)$  and  $\sigma_2^2 = (1 - \beta)R/P_c$ .

#### Algorithm 1 Asynchronous Best-Response Algorithm

Output:  $j, j \in \{e, c\}$ 

**Input:** Initialize k as 1 and randomly pick a feasible  $P_j^{(0)}$ 

- 1: for iteration k do
- 2: Receive the miners' request vectors  $\mathbf{r}^{(k-1)}$
- Predict the strategy of the other SP
- 4: Decide  $P_j^{(k)} = P_j^{(k-1)} + \Delta \frac{\partial V_j \left( P_j, P_{-j}^{(k-1)}, \mathbf{r}^{(k-1)} \right)}{\partial P_j}$
- 5: **if**  $P_j^{(k)} = P_j^{(k-1)}$  **then** Stop
- 6: **else** send  $P_j^{(k)}$  to miners and set  $k \leftarrow k+1$

上层解法: Asynchronous Best-response Algorithm





**Theorem 3.** The unique Nash equilibrium for miner  $m_i$  in the NEP<sub>HOMOMINER</sub> problem is given below

$$\begin{cases} e_i^* = \frac{B\beta h}{(1-\beta+h\beta)(P_e - P_c)}, \\ c_i^* = \frac{B[(1-\beta)(P_e - P_c) - P_c\beta h]}{P_c(1-\beta+h\beta)(P_e - P_c)}, \end{cases}$$
(28)

provided that the prices set by the ESP and the CSP satisfy  $P_c < \frac{1-\beta}{1-\beta+h\beta}P_e$ .

#### 同质矿工的情况下,可以轻易得到下层解

Problem 2b (SP SUBGAME: NEPSPHOMOMINER).

maximize 
$$V_e = (P_e - C_e) \cdot e_i^*, V_c = (P_c - C_c) \cdot c_i^*,$$
 (30a)

subject to 
$$P_c < \frac{1-\beta}{1-(1-h)\beta}P_e$$
, (30b)

where 
$$e_i^* = \frac{B\beta h}{(1-\beta+h\beta)(P_e-P_c)}$$
,  $c_i^* = \frac{B[(1-\beta)(P_e-P_c)+P_c\beta h]}{P_c(1-\beta+h\beta)(P_e-P_c)}$ .

**Theorem 4.** The unique Nash equilibrium for the SPs in the NEP<sub>SPHOMOMINER</sub> problem is given below:

$$\begin{cases}
P_e^* = \bar{p}, \\
P_c^* = \frac{C_c \bar{p} (1-\beta) - \bar{p} \sqrt{C_c h \beta (\bar{p} - C_c)(1-\beta)}}{[1-\beta(1-h)]C_c - \beta h P_e},
\end{cases} (31)$$

where  $\bar{p}$  is the solution to  $\partial V_e/\partial P_e = 0$ .

### 上层解也可以求得

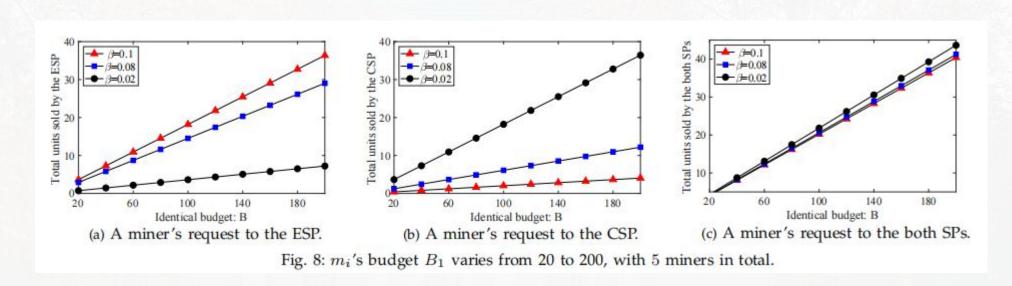


实验步骤









### 不同预算下SP的收益

止于至善 18





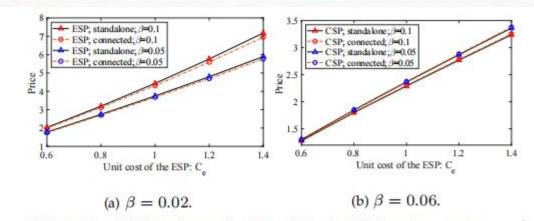


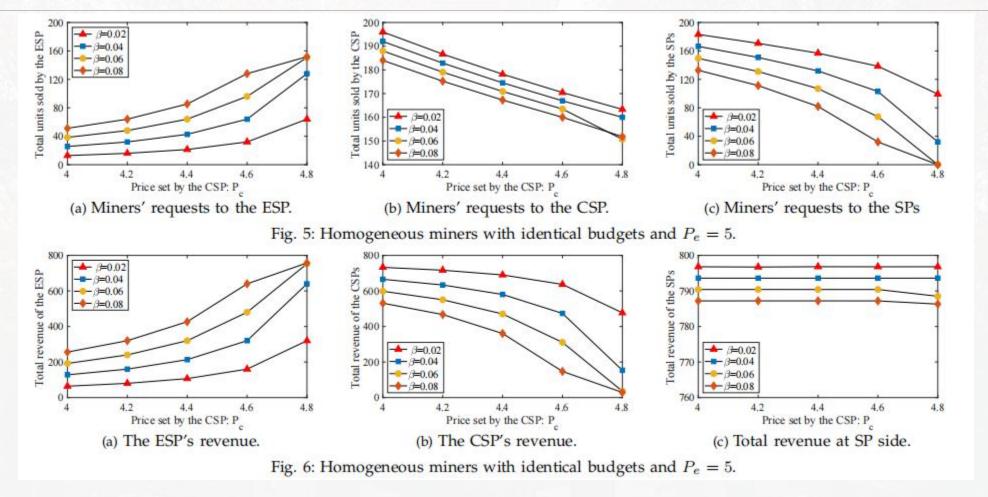
Fig. 9: The CSP's unit cost is 0.5, while the ESP's unit cost changes.

随ESP的cost的变化,其价格的变化



# 同质矿工实验





### 同质矿工的实验结果图

业于至善20





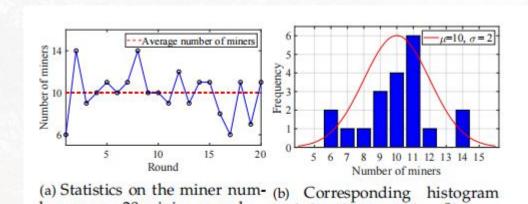
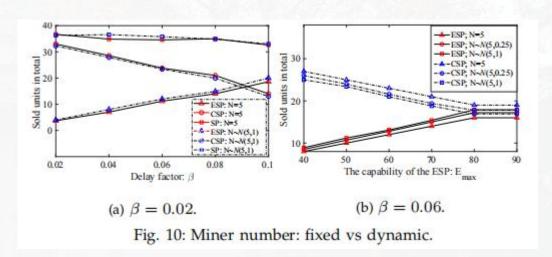


Fig. 4: A toy example for population dynamics of moblie miners. TABLE 2: Optimal requests of homogeneous miners with sufficiently large budgets where  $\gamma = (N-1)R/N$ .

ber among 20 mining rounds. and distribution  $N(\mu, \sigma^2)$ .

### 矿工数服从N(10,4)分布



#### 动态静态对比图



# 结论

Research Methods And Processes







#### ■ 本文总结

- 我们提出了一个最优价格策略和最优计算卸载请求的移动 矿工之间的斯塔克尔伯格博弈
- 讨论了该对策中斯塔克尔伯格平衡的存在性和唯一性,并 提出了实现SE点的算法
- 不确定性使矿工更积极地从ESP购买计算资源

#### ■ 留下的思考

- 模型设计还可以继续优化
- 上层的均衡证明比较模糊
- 依旧是通过迭代进行数值求解



# 感谢大家的观看

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