

Curated Proof Markets & Token-Curated Identities in Ocean Protocol

Trent McConaghy
@trentmc0



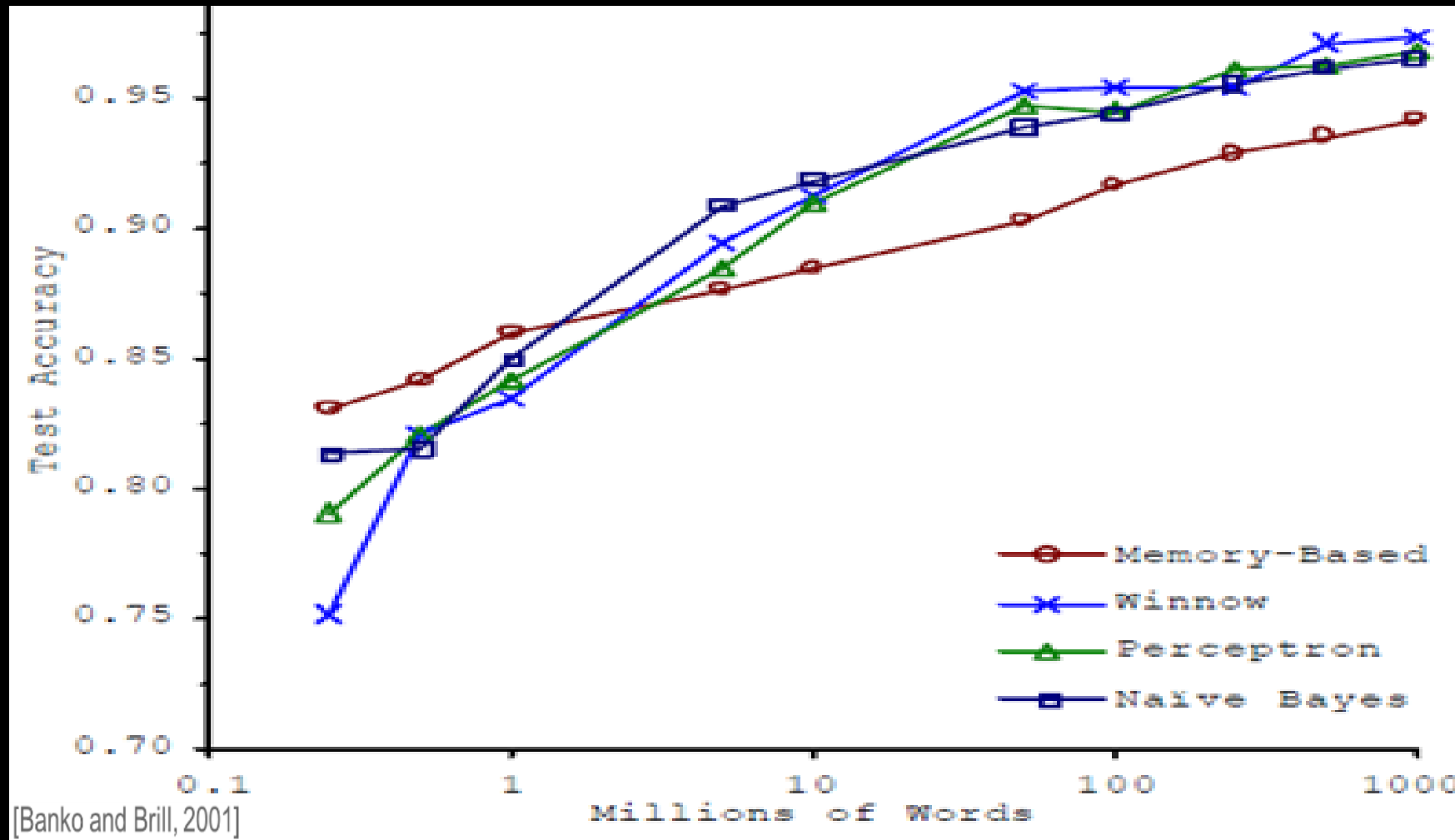
#Data
#Incentives





Audio radar

The Unreasonable Effectiveness of Data



1000%
less
error!

1000x *more* data

The world's most valuable resource



Data and the new rules
of competition

Silo mo' data



Mo' accuracy



Mo' \$

Default incentive:
hoard the data

**“Show me the incentive
and I will show you the outcome.”**

-Charlie Munger

You can get people to do stuff
by rewarding them with tokens.
This is a superpower.





Change the
incentives!

~~Site~~ *Pool* mo' data



Mo' accuracy



Mo' \$

How to design?

Here's a Process for Token Engineering

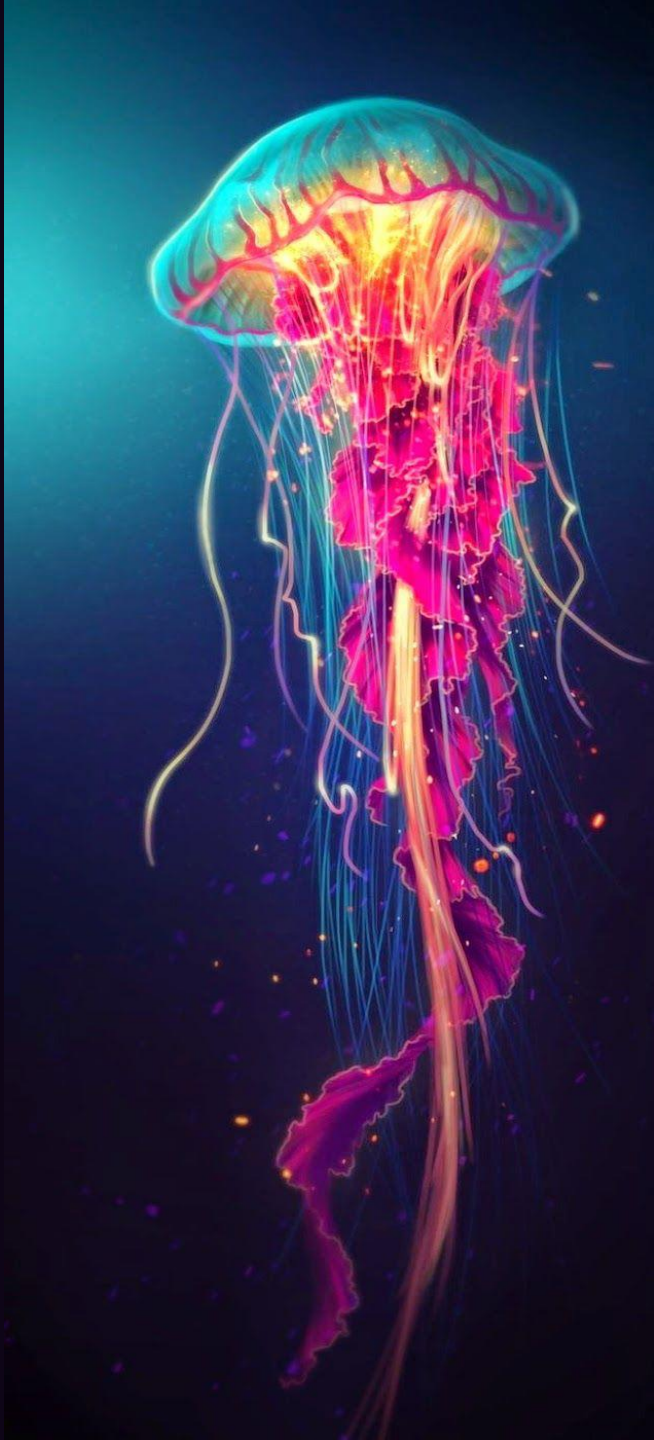
1. **Formulate the problem.** Objectives, constraints, design space.
2. **Try an existing pattern.** If needed, try different formulations or solvers.
3. **Design new pattern?**

Patterns (Building Blocks) for Token Design

- Curation
 - Stake machines, curation markets
 - Curated proofs markets, TCRs
- Proofs
 - Bitcoin puzzle solving, Filecoin PoST, Steemit human
 - Data availability, more
- Identity
- Reputation
- Governance / software updates
- Third-party arbitration
- ...

Patterns (Building Blocks) for Token Design

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Curated Proofs Markets: Formulation

Objective: maximize supply of relevant data

- We convert this objective to block rewards.
- Then, how about this reward:
“Reward making relevant data available when asked”
- But: How to know what’s relevant? Algorithms capture this poorly.
- Solution: leave it to the crowd. Let them put their money where their mouth is! (Staking)
- Revised reward:
“Reward curating data (staking on it) + making it available”

Objective: maximize supply of relevant data

- Reward curating data (staking on it) + making it available
- New pattern: curated proofs market

$$E(R_{ij}) \propto \log_{10}(S_{ij}) * \log_{10}(D_j) * T * R_i$$

Expected
reward for user
 i on dataset j

S_{ij} = predicted popularity
= user's curation market
stake in dataset j

D_j = proofed popularity
= # times made dataset
available

tokens
during
interval

From AI data to AI *services*

Motivations:

- Privacy, so compute on-premise or decentralized
- Data is heavy, so compute on-premise
- Link in emerging decentralized AI compute

Objective function: Maximize supply of relevant *services*

=reward curating *services* + proving that it was delivered

$$E(R_{ij}) \propto \underbrace{\log_{10}(S_{ij})}_{\text{predicted popularity of service}} * \underbrace{\log_{10}(D_j)}_{\text{proofed popularity of service}} * T * R_i$$

$$E(R_{ij}) \propto \log_{10}(S_{ij}) * \log_{10}(D_j) * T * R_i$$

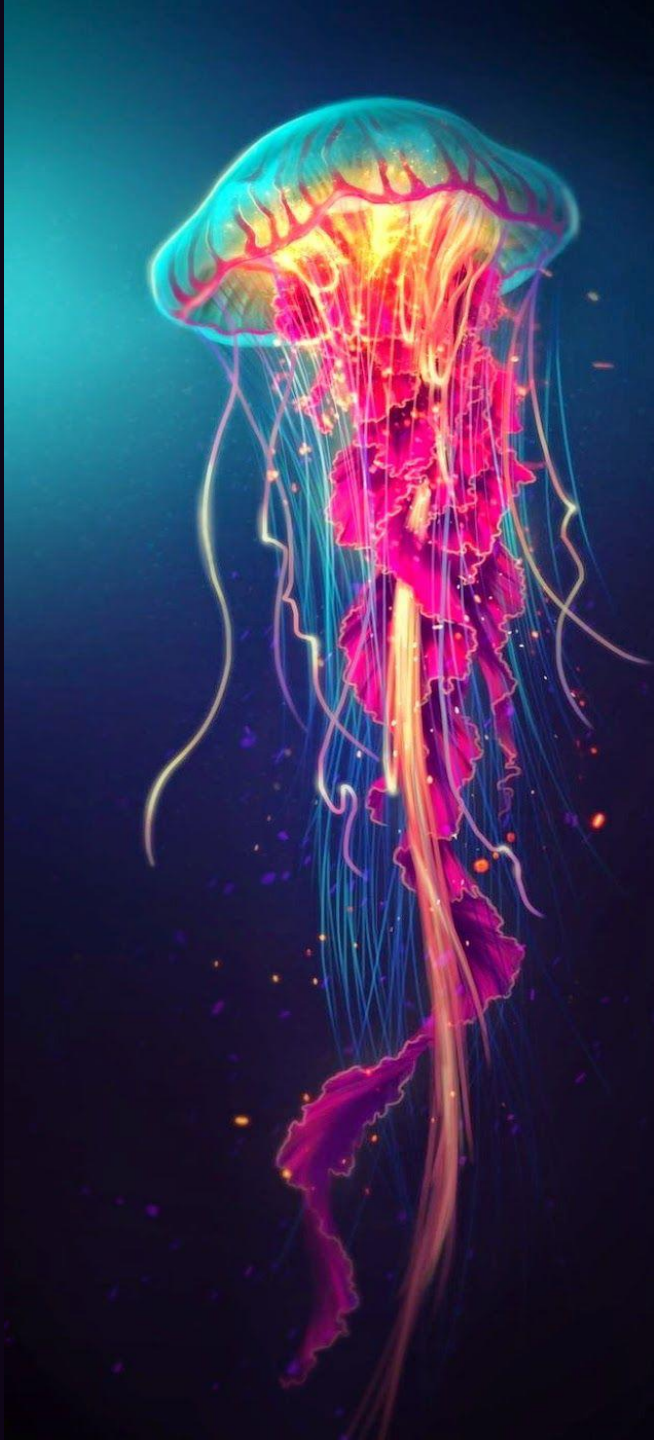
S_{ij} = expected popularity =
stake in dataset j

Key block #1:
Curation market

D_j = actual popularity = #
downloads

Key block #2:
Proof of availability

(All other blocks are ancillary to these blocks)



Proofs

$$E(R_{ij}) \propto \log_{10}(S_{ij}) * \log_{10}(D_j) * T * R_i$$

S_{ij} = predicted popularity =
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Key block #1:
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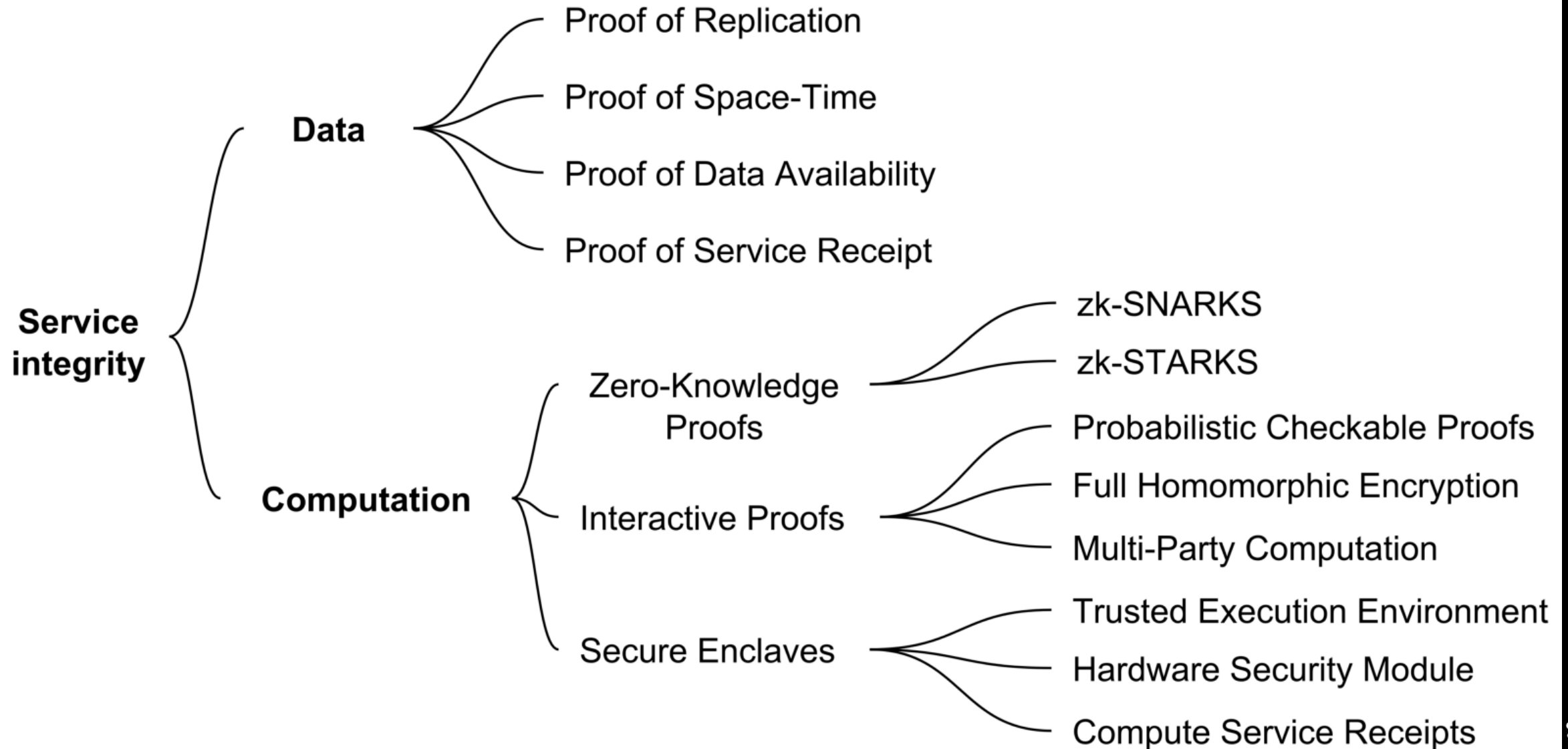
Proof of Data Availability

Approach A: FileCoin Proof of Space-Time (PoST), though that's overkill

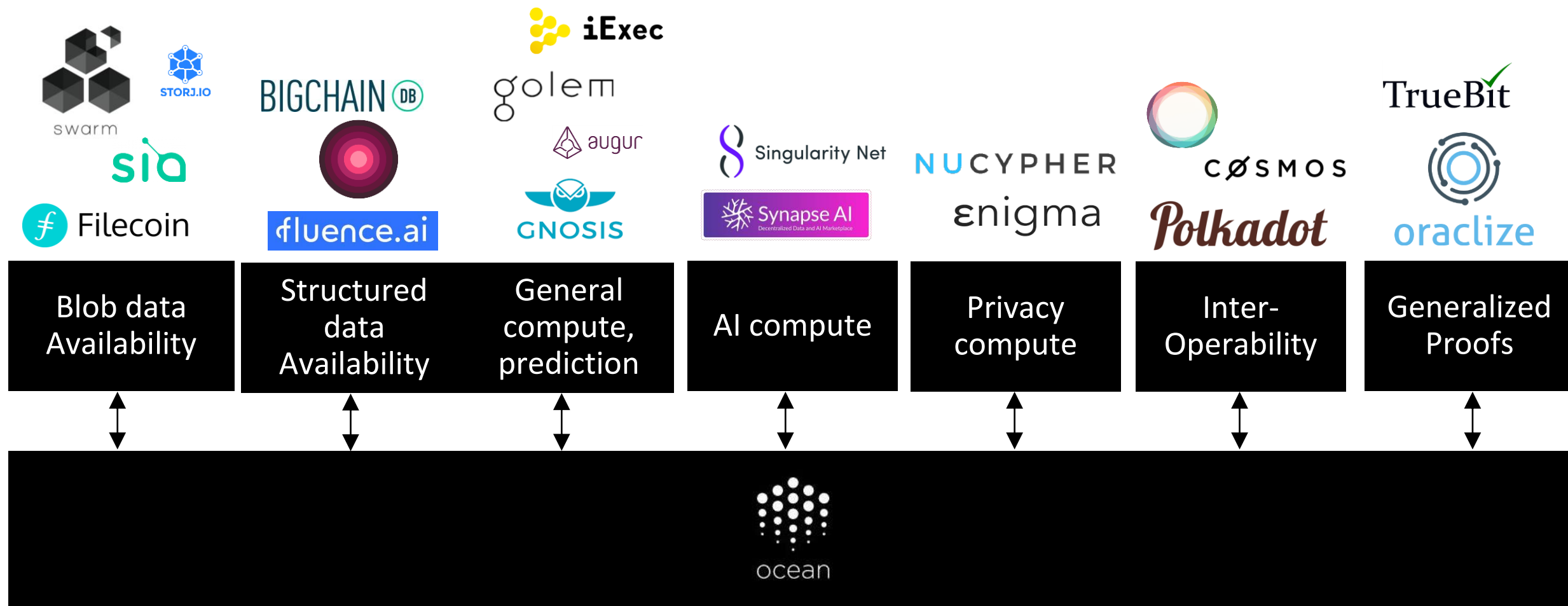
Approach B: TrueBit-style challenge-response:

1. I serve data to you
2. If you don't think you got it, you challenge (with stake)
 1. Randomly choose two actors to vet
 2. Actors vote on whether served
 3. If they agree that I served it, you lose stake
 4. If they agree that I didn't serve it , I lose stake
 5. Else

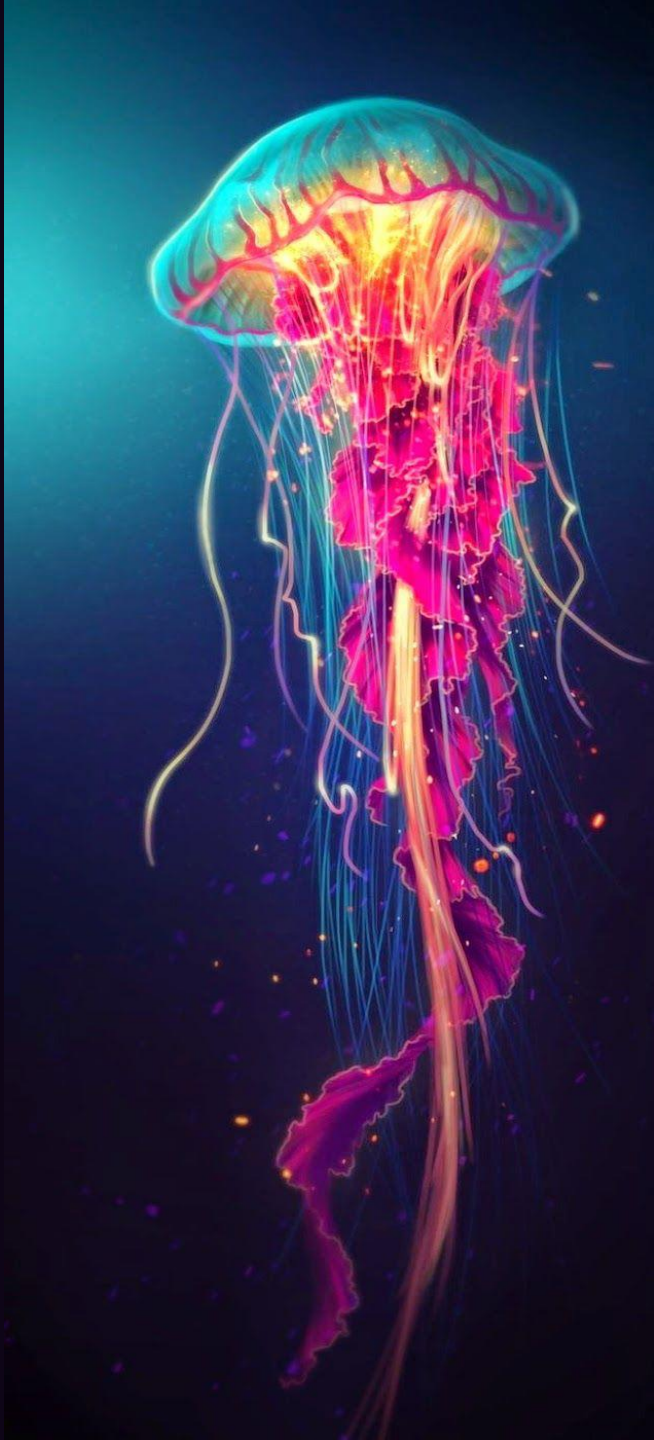
More Proofs



With more proofs, more services!



*Note: logos shown are examples and do not imply partnerships or integrations



Token Curated Registry (TCR)

First, some attack vectors



Settings Tools

View saved SafeSearch ▼

This is a large, dense collage of numerous small images related to Disney's Frozen franchise. The images include: Elsa in her ice queen attire; Anna in her red dress; Olaf the snowman; Kristoff and Sven; Hans; and other characters like Yelena Belova from the animated short "Frozen Fever". There are also fan art pieces, character comparisons (e.g., Elsa as Spider-Man), and scenes from the movie itself. The collage is organized into several rows, with some images being larger than others, creating a visually busy and comprehensive collection of Frozen-related content.

Some attack vectors

- **Elsa & Anna Attack:** Someone uploads popular content that they don't have rights to
- **Data Escapes:** People take data out of the system
- **Curation Clones:** Others create a new market for an existing dataset
- **Sybil Downloads:** Someone (or a ring of buddies) downloads a dataset they own $\gg 1$ times, to get \gg rewards

How to approach?

Idea 1: network directly has legal arbitration

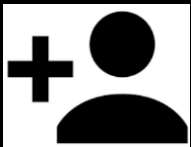
- Problem: it binds the network to jurisdictions (eek).
- *Need* to be permissionless and borderless!!

Idea 2: “sunny day tech, rainy day law”

- Core network itself has self-regulation by community.
Incentivize good acting. Zero sum → positive sum. 99% case.
- Higher levels can have arbitration tied to jurisdictions. 1% case

TCR: a whitelist of good actors, curated by the actors themselves

New user B
proposes
for B to join



10 tokens
stake
by B

```
class TokenCuratedRegistry
  def propose(data)
  def challenge(proposal)
  def vote(challenge)
```

User	status
A (Alice)	"Challenger"
B (Bob)	"New" + "Proposed"
Mallory	"OK"
Trent	"OK"

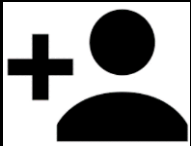
Existing user A
challenges

10 tokens
stake
by A



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10 tokens

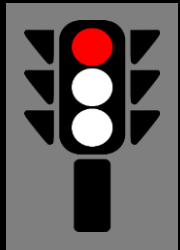
```
class TokenCuratedRegistry
  def propose(data)
```

Concern:

**High friction onboarding
7-28 day waiting period**

Existing user A
challenges

10 tokens



A (Alice)	"Challenger"
B (Bob)	"New" + "Proposed"
Mallory	"OK"
Trent	"OK"

TCR: a whitelist of good actors

“Trust is risk” for low-friction onboarding: vouch for others

Existing user
A *vouches*
for new user
B to join



10 tokens
stake by A,
B can run
away with it
anytime

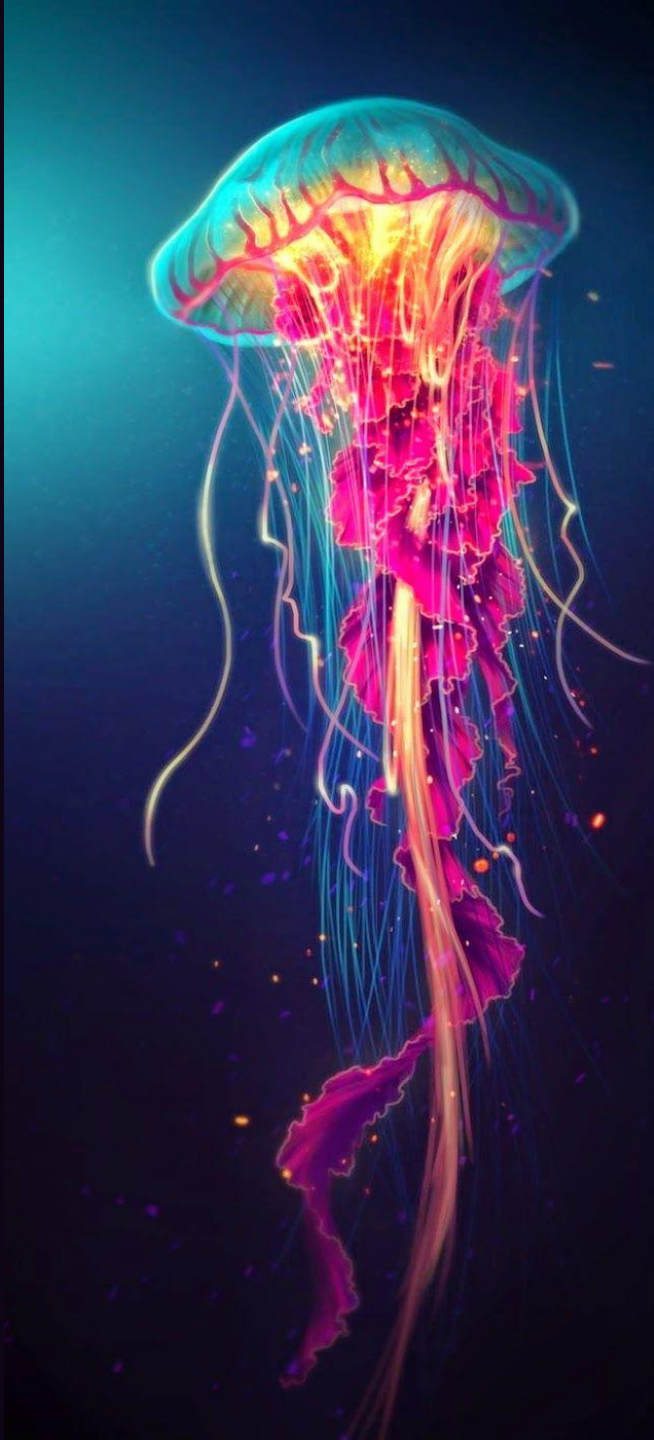
```
class TokenCuratedRegistry
  def vouch(data)
  def challenge(proposal)
  def vote(challenge)
```

User	status
A (Alice)	“Vouches”
B (Bob)	“New”
Mallory	“OK”
Trent	“OK”

Some attack vectors

- **Data Escapes:** People take data out of the system
- **Curation Clones:** Others create a new market for an existing dataset
- **Elsa & Anna Attack:** Someone uploads popular content that they don't have rights to
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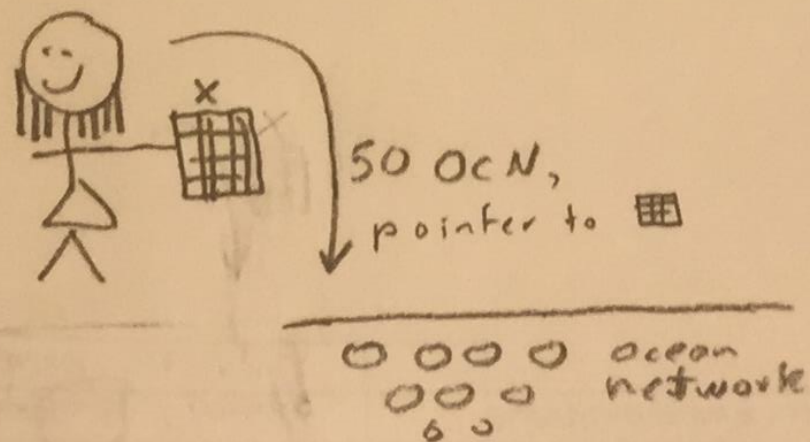
Whitelist of good actors addresses all of these!



Walk Through

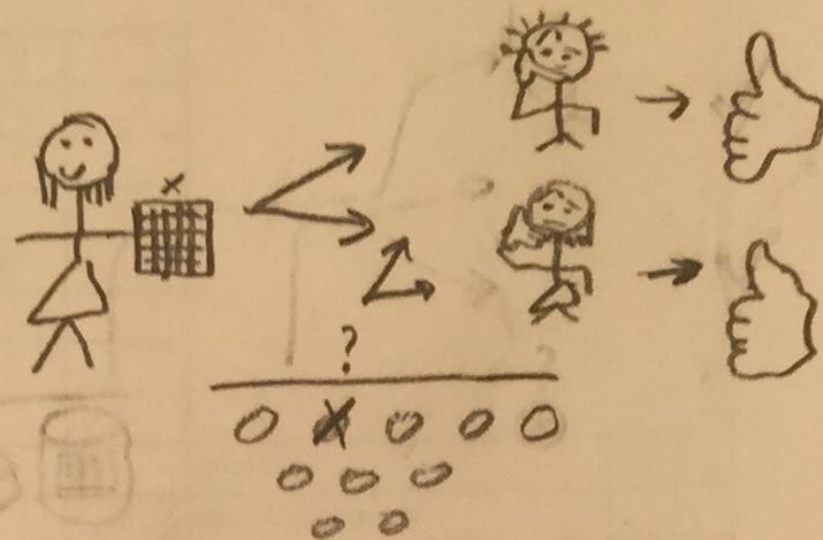
[Claim a dataset]

①



Alice claims copyright on dataset X.
and stakes 50 ocN to do so.

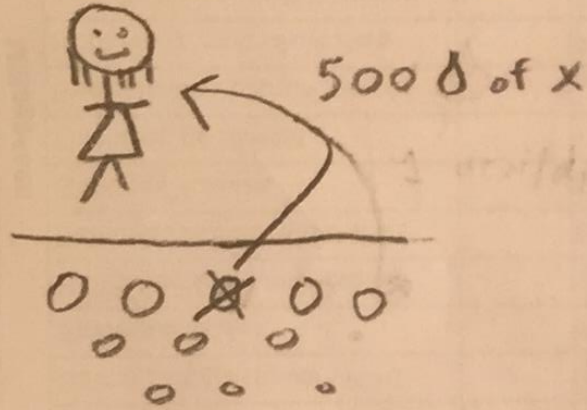
②



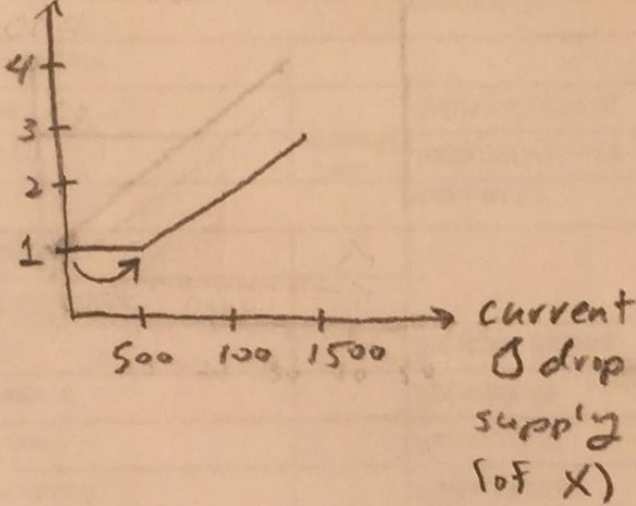
Others check. Data is ok!
5 day period concludes.

[Initiate curation, invest more]

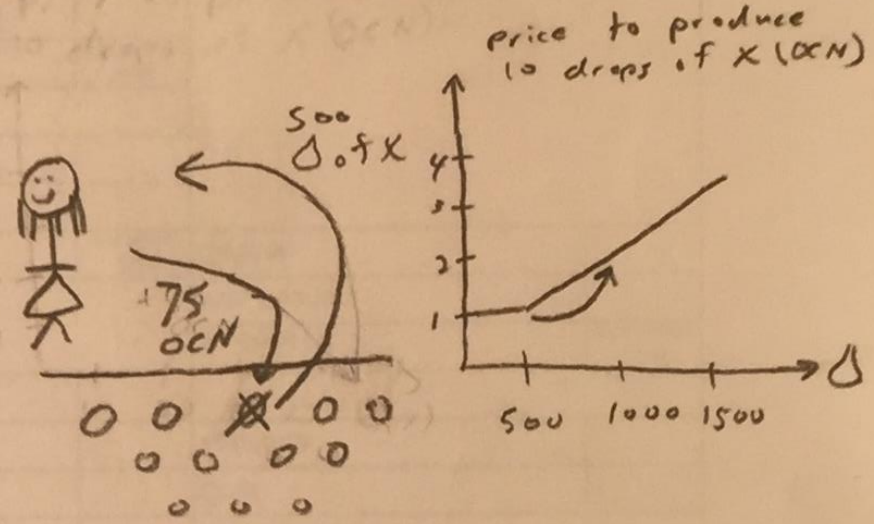
①



Price to produce
10 drops of X (OCN)



②



The curation market for X is initialized,
with an initial supply of 0 drops.

At that supply, 10 drops cost 1 OCN.

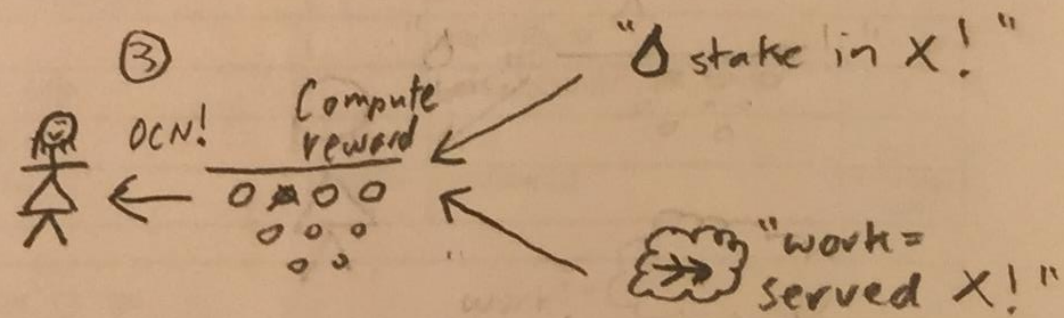
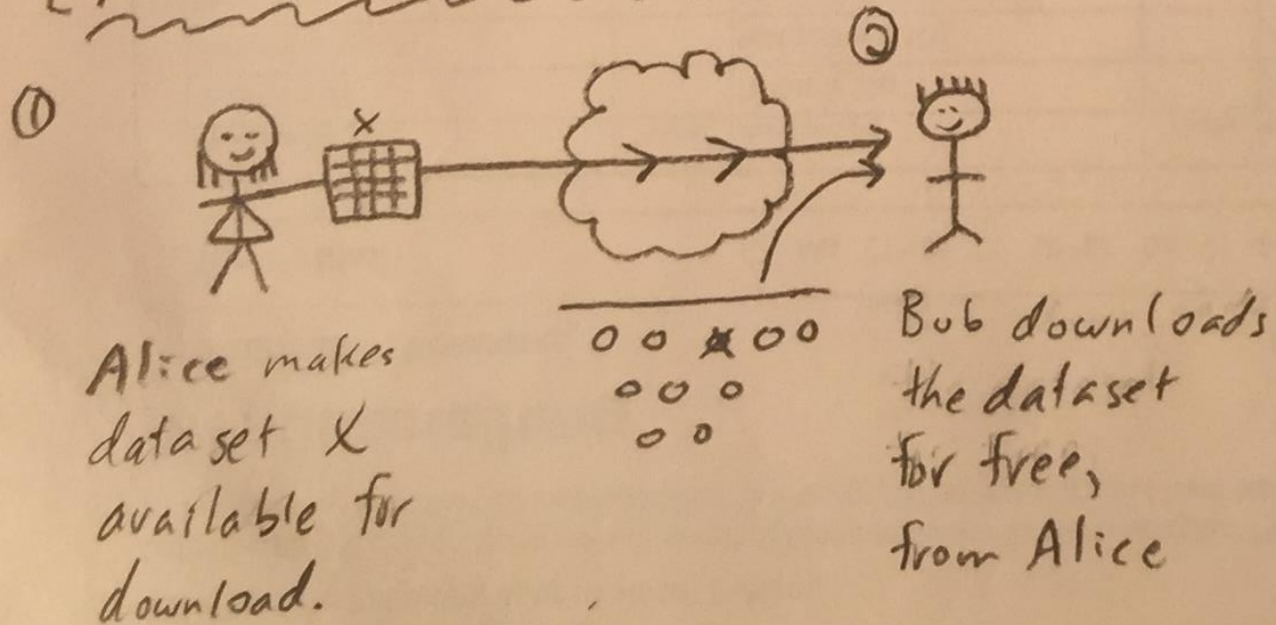
Because Alice staked 50 OCN, she
gets $10 \cdot 50 = 500$ drops of X.

Drops have value for block rewards and
when un-staking, as we will see.

Alice has high confidence
in the future popularity
of X. So, she gets

another 500 drops of
X by staking $\frac{(1+2)}{2} \cdot 50 = 75$
OCN.

[Make data available, block rewards I]

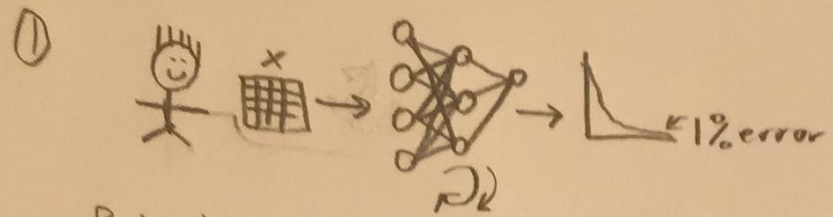


Alice gets block rewards (on average) because she's staked on X, and served it when requested.

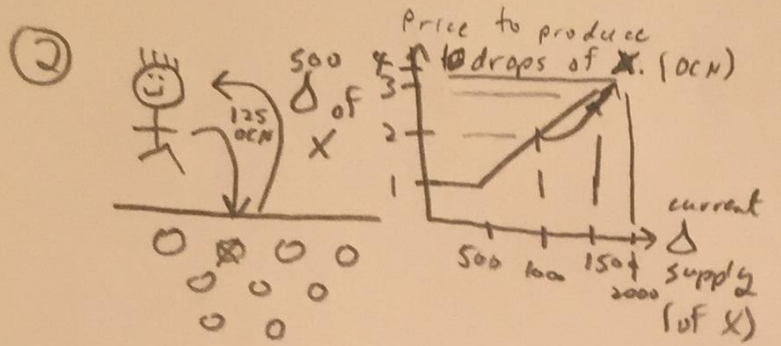
$$\text{Reward} \approx \frac{\text{Alice's } \Delta \text{ stake in X}}{\text{difficulty}^*} = \frac{1000}{\text{diff.}}$$

* details in whitepaper

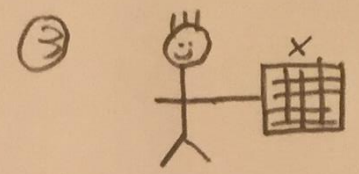
[Referral / curation]



Bob tries the dataset out on his AI model, and finds that it's really useful. Cool!



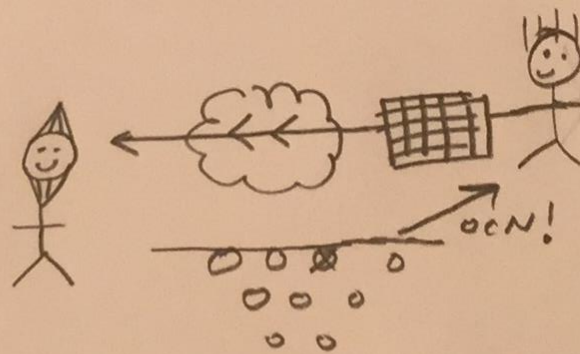
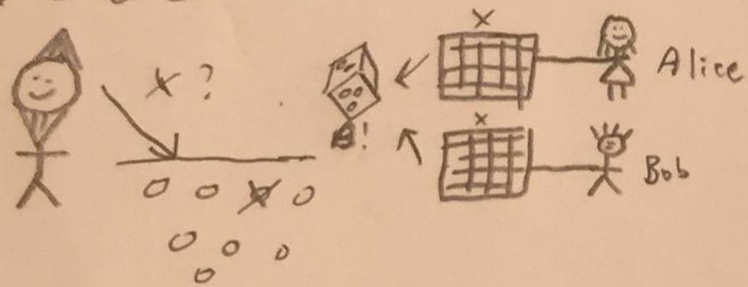
Bob realizes that others will find the dataset useful too. So, he gets 1000 drops of X by stating an average price of $\left(\frac{2+4}{2}\right) = 3 \text{ OCN}$ for every 10 drops of X. So he spends 300 OCN for 1000 drops.



Bob makes dataset X available for download too.

$$\frac{3 \text{ OCN}}{10 \text{ drops}} = \frac{300}{1000}$$

[Block rewards II]



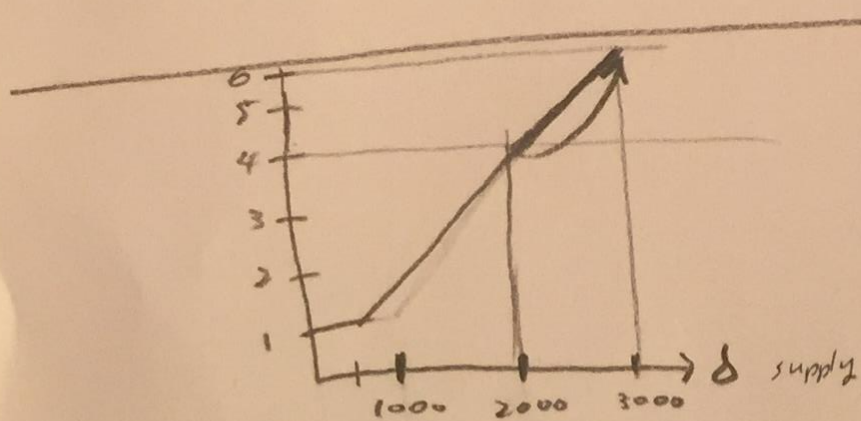
① Joe requests dataset X.

② From providers, Bob is randomly chosen

③ Joe downloads X from Bob

④ Bob gets block rewards (on average) because he staked on X, and served it when requested.

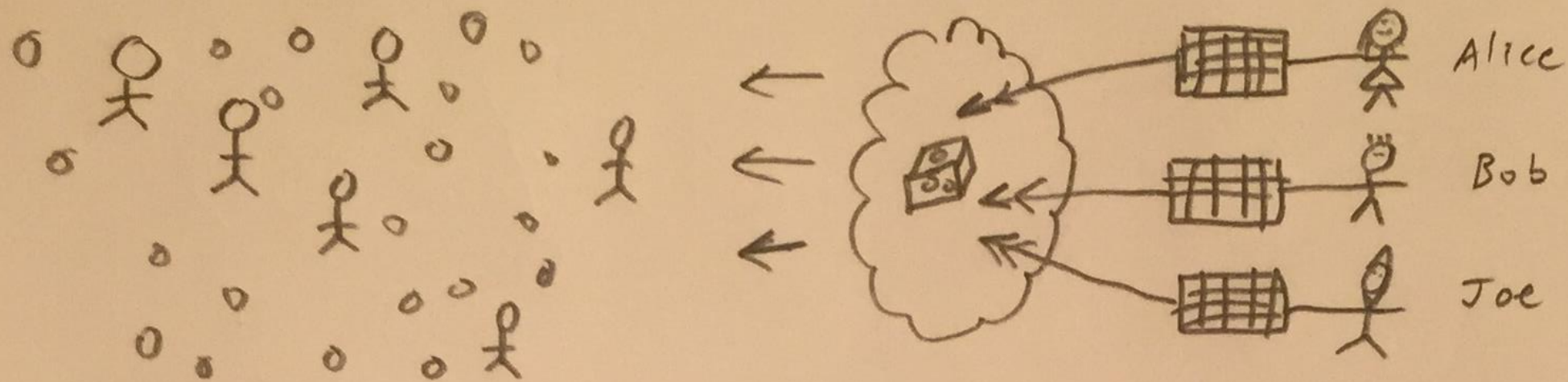
$$\text{Reward} \approx \frac{\text{Bob's stake in X}}{\text{difficulty}} = \frac{1000}{\text{diff.}}$$



⑤ Joe likes the dataset too, he believes in its future popularity, so he buys 1000 drops of X, at an average price of $\left(\frac{4+6}{2}\right) = 5$ oCN / 10 drops, or 500 oCN for 1000 drops.

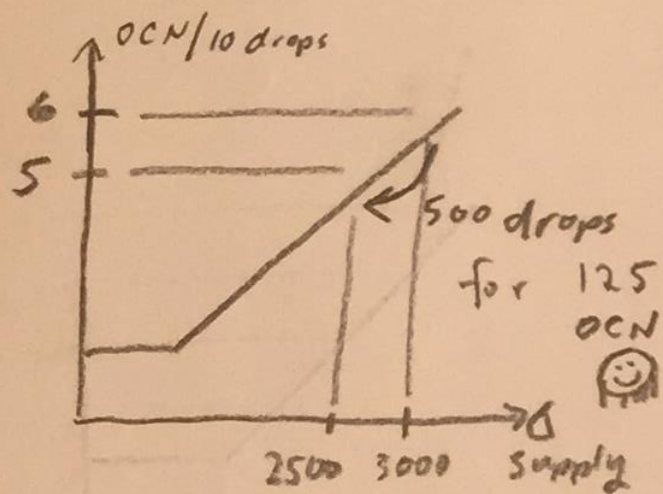
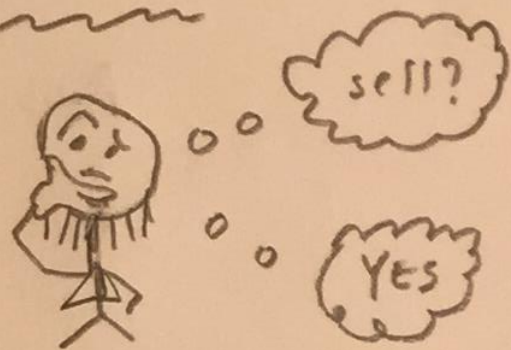
Note how Alice and Bob get the same reward on average because they both have 1000 δ . However, Alice staked 125 oCN versus 300 for Bob, because she was an earlier adopter!

[Block rewards III: payoff on average]



Another 100 people download the dataset. It's becoming popular!
Alice, Bob, and Joe each serve up \times approximately $\frac{1}{3}$ the time,
and get equal rewards (because equal δ).

[Sell stake I]



Alice considers selling 500 Δ of X.

Supply is currently 3000 Δ ,

therefore average price is $(\frac{5+6}{2}) = 5.5$ OCN / 10 drops.

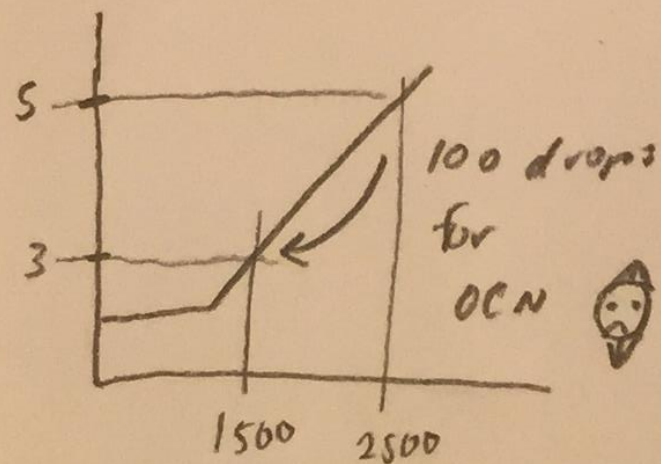
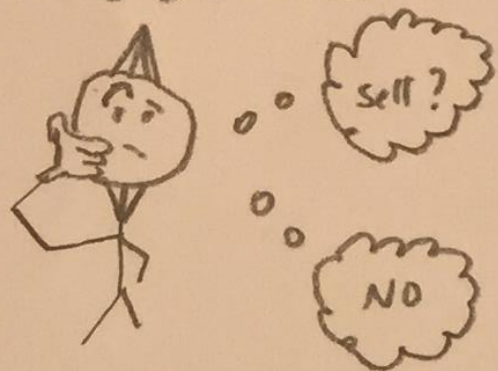
She sells, getting $(5.5) \cdot (50) = 275$ OCN.

Wow! She originally invested $50 + 75 = 125$ OCN,
and she's already got that back, and more!

Plus she's retained half her Δ .

Early adopter FTW!

[Sell stake II]



Joe considers selling 1000 S.

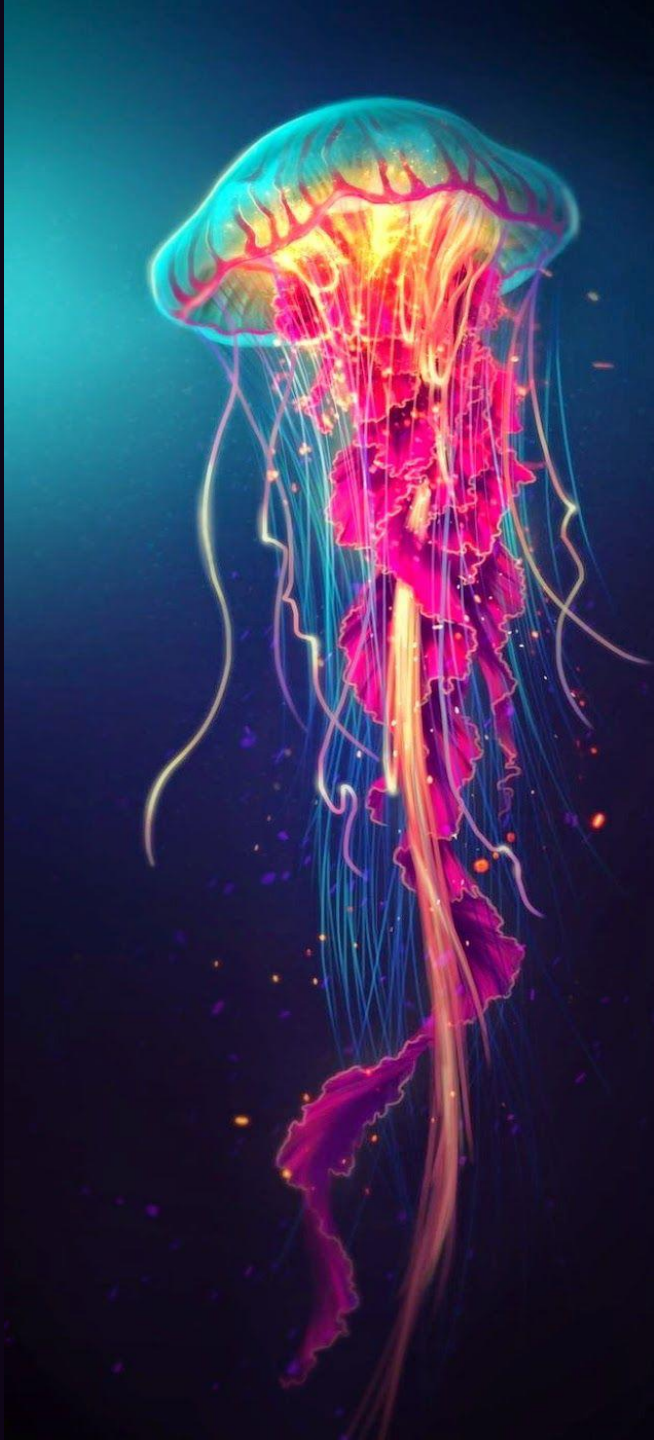
Average price is $\left(\frac{5+3}{2}\right) = 4$ OCN/10 drops.

He'd get $(4)(100) = 400$ OCN.

Recall that he spent 5 OCN/10 drops, or 500 OCN for 1000.

So, he'd lose money. He doesn't sell. He's a late adopter.

Instead, he waits; he thinks others might invest.



Conclusion

AI → Data silos → data crisis.

**A thoughtful token design
has thoughtful building blocks.**

For Ocean, this includes:

- **Curated Proofs Markets (CPMs)**
 - **Token Curated Registry**