On Decentralizing Al Data & Services

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#Data #Incentives



Audio radar

The Unreasonable Effectiveness of Data



1000x more data



Silo ++ data ++ accuracy ++ \$

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!

Sile Pool ++ data ++ accuracy ++ \$



Early iterations

X=[]; y=[] (X=C]; y=C] neuX=50 initial samples in [G1] Jsome based on Jumerai Jumerai For each x: ([1]) in new X Proposterscated data: given D1, D2,... (date set 1, 2...) create O: Sobfiscated clata based on 2. e.g. $Q_{i} = \xi D_{2}, D_{3}$ for x = [0, 1, 1, 0]Stor each data serentist For subset of neux (he choosed) it Build model mapping of strade sh Commit stake. It's a proxy for Decte. scientisi Submit model + stake to Numera, + data, and pages for compute cost Numeria Deploy all trading models. Wait 30 d. For each ki in new X freevidit Ecompute & from traday = yi. Update y; in nuly i.e. pp X=XU new Y: y=YU new Y usm Build model if X=1 (ditachoices-walke) data Computer impacts of each imput in model (ie data set paging Compute pagoat to data provides (upute pagodt. go screeting aller the der. proc. (smert contrad) und chose new dir yard on mades aller screetings Aut = choose new dir yard on mades aller screetings low firther price -homomorphic ener. Les cost Czkp to trustless computing 201 o high friction sabaril - intel /ms secure enclaves low triction price optimally fair /transport - latert variable obtassation (GANS





Multi-level a action, supplier set price Sata bank Thave IK miles , bidden plane XBOX TOP 2. Auction happens are -sof conditions based on the bid, when you get to as it #2 500) highest > \$ 10K now TOP 3 in 2 ms 16K Ind-highesthids got data \$3K in 2 mu \$IK now"; in 3 m. rest in 1-6 mos" And : in 6 mos: data is set free. Stermit /MCM - post content, - post data, got takens got duters - spond taken to upwate - it others we wate; 3 1. Supplier says: total price \$LOK. Tap biddless in pool get data now " De Fascaled later 13 only winkle by Marketplace for obtescalis data Numerar itself" Pooling with Tap biddens in pool get Rest get data later. Supplior-set price these folks took set supple. IOK 7K \$ 32K zon 1× 2× others Set data in Aller is it and (idete sty Free o)

Engagements / Incentives & marbets no into do por ever data - & distance fix price up-front 11 reputation (subjective) on-data 1-You have SOK stated in - 1/10,000 we lid Supplier scientist XA -ead key is a percention hash of a row of resta. (HD willet) 2. All to compute a key from data is public Stake 3. .. fory data is made public, then Anyone can open my wallet leget \$50k. model poisnes visible for data supplies data visible for scientist data puiste - madel parans private public ouf. printe Imodel use obtinuted public toten dist. or Here Markets? date data co-ourer of \$ 1 prodel model Takensing Access to Data Revenue Fixed surgery - Each defect has its own tilms. Pixen sweet - "ICO a dataset" - WICO a dataset" - Wich defa is particular for usess, \$ is split necerds to At 1-04 token ownership Tokenizing Access to Data Itzelf -e.g. 100 tokens. You can access the date if you own the token. 00

Challenges O. Denvertix late. 1. How to ensure supplier get pail who losing ability to get paid in Eutome. "Free riding" "Privacy" "Copy is title" 2. Friction in pricing) - percelative impact per dataset Static -> Dynamic dataset Fring ilder Non Fring ilde (gars stale) - data labeling source - data obfuseation market Signals Tools to address file ridery - Stake in bolief of take working value of dataset States - Set the Frer " after De Detring to - Data provider doesn't care - by lata scialit - Licensing - Aron gin - GURR - price asked by supplies readine - price bid by scientist - reputation e-reputation of dataset - provenance -visk of litigate - supplier - only the smart contract can see - -- scientist the data. Eg docker + locks - bandies for non-freed viding detection P Svalac gained from dalasel (s) in Pet FL F-tatal value at nature a view for and strate - prediction market belief in value of -If data set free, for the your muste key gots exposed my stake of ### date at - novelty of a dataset -Data obtescation, as latent variables on NN (like take a price of data set Narmoni malimite Tay GANSI



Early iterations: Flailing

Can we structure this better?

Realization: Tokenized Ecosystems Are a Lot Like Evolutionary Algorithms (EAs)!

What	Tokenized ecosystem	Evolutionary Algorithm
Goals	Block reward function E.g. "Maximize hash rate"	Objective function E.g. "Minimize error"
Measurement & test	Proof E.g. "Proof of Work"	Evaluate fitness E.g. "Simulate circuit"
System agents	Miners & token holders (humans) In a network	Individuals (computer agents) In a population
System clock	Block reward interval	Generation
Incentives & Disincentives	You can't control human, Just reward: give tokens And punish: slash stake	You can't control individual, Just reward: reproduce And punish: kill

We can approach token design as EA design.

EA Design



Steps in EA (and Optimization) Design

1. Formulate the problem. Objectives, constraints, design space.

2. Try an existing solver. If needed, try different problem formulations or solvers.

3. Design new solver?

1. Formulation of an optimization problem Objectives & constraints in a design space

The algorithm's aim is formulated as a constrained multiobjective optimization problem

minimize
$$f_i(\phi)$$
 $i = 1...N_f$
s.t. $g_j(\phi) \le 0$ $j = 1...N_g$
 $h_k(\phi) = 0$ $k = 1...N_h$
 $\phi \in \Phi$
(1)

where Φ is the "general" space of possible topologies and sizings. The algorithm traverses Φ to return a Pareto-optimal

2. Try an existing solver. Does it converge?



3. Design new solver

e homo-	TABLE II
motopy	PROCEDURE SANGRIAOPTIMIZATION()
coarsely	Inputs: $D, N_a, K, N_L(k)$
ructural	Outputs: d^*
y. Tradi-	1. $N_{gen} = 0; P = \emptyset, P_{all} = \emptyset$ 2. while stop() $\neq True$:
ro path,	3. if $(N_{gen} \% N_a) = 0$:
the zero	4. if $ P < K$: 5. $P_{ P < I} = \emptyset$
several	6. $P_0 = $ SpaceFillIndividuals $(N_L(k), N_D, D)$
	7. for $k = 1$ to $ P $:
mulated	8. $P_k = \text{SelectParents}(P_k, P_{k-1}, N_L(k))$ $P_k = \text{Lindetal acclOntState}(P_k, P_{k-1}, N_L(k))$
nalyses,	9. $P_{k,j} = \text{OpdateLocalOpiState}(P_{k,j}, \kappa), j = 1 \text{ to } P_k $ 10. $P_{all} = \text{unique}(P_{all} \cup P)$
oint $\boldsymbol{\theta}$.	11. $P_{ P } = P_{ P } \cup \text{InnerOptimize}(P_{all}, D, k)$
nt/other	12. $d^* = d_i$ in P_{all} with highest Y or Cpk
onnom-	13. $N_{gen} = N_{gen} + 1$ 14. return d^*
corners	
rated in	and all individuals encountered so far in the search P_{-11}
on (with	Lines 2 12 are the generational loop, which repeats until stop

Example of a Successful Outcome



Token Design as EA Design



Steps in Token Design

1. Formulate the problem. Objectives, constraints, design space.

2. Try an existing pattern. If needed, try different formulations or patterns.

3. Design new pattern?

1. Formulate the Problem

(a) Ask

• Who are my potential stakeholders?

- And what do each of them want?
- What are possible attack vectors?

(b) Translate those into objectives and constraints.

2. Try Existing Patterns

- 1. Curation
- 2. Proofs of human or compute work
- 3. Identity
- 4. Reputation
- 5. Governance / software updates
- 6. Third-party arbitration
- 7. ...

2.1 Patterns for Curation

- **Binary** membership: Token Curated Registry (TCR)
- Discrete-valued membership: Stake Machines
- Continuous-valued membership: Curation Markets characterized by bonding curve
- Hierarchical membership: each label gets a TCR
- Work tied to membership: Proofed Curation Market
- Non-fungible tokens: Re-Fungible Tokens



Case Study: Analysis of Bitcoin

Bitcoin objective function



Objective: Maximize security of network

- Where "security" = compute power
- Therefore, super expensive to roll back changes to the transaction log

Bitcoin objective function

Objective: Maximize security of network

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Result of Bitcoin's objective function: People are maximizing security! = Maximizing electricity



Case Study: Design of Ocean

Formulate the Problem: (a) Who are stakeholders? What do they want?

Key stakeholders in Ocean ecosystem								
Stakeholder	What value they can provide	What they might get in return						
Data/service provider, data custodian, data owner	Data/service (market's supply)	Tokens for making available / providing service						
Data/service referrers, curators. Includes exchanges and other application-layer providers.	Data/service (via a provider etc), curation	Tokens for curating						
Data/service verifier. Includes resolution of linked proofs on other chains	Data/service (via a provider etc), verification	Tokens for verification						
Data/service consumer	Tokens	Data/service (market's demand)						
Keepers	Correctly run nodes in network	Tokens for chainkeeping						

Formulate the problem: (b) Translate into objectives and constraints

Objective function: maximize supply of relevant data

Token rewards if: supply relevant data Token rewards if: supply data, and curate it

Formulate the problem: (b) Translate into objectives & constraints

Constraints = checklist:

- For priced data, is there incentive for supplying more? Referring?
- For priced data, good spam prevention?
- For free data, is there incentive for supplying more? Referring?
- For free data, good spam prevention?
- Does the token give higher marginal value to users of the network versus external investors? Eg Does return on capital increase as stake increases?
- Are people incentivized to run keepers?
- Is it simple? Is onboarding low-friction?

2. Try Existing Patterns Some patterns:

- 1. Actor registry
- 2. Data registry
- 3. Actor registry + data registry
- 4. Data registry + free-as-in-beer data curation market. Curation: Pay tokens to listen.

2. Try existing patterns: evaluate on objectives & constraints. None passed...

Key Question	1	2	3	4
For priced data: incentive for supplying more? Referring?	×	*	✓	*
For priced data: good spam prevention?	*	~	~	~
For free data: incentive for supplying more? Referring?	×	~	×	~
For free data: good spam prevention?	*	~	*	~
Does token give higher marginal value to users of the network, vs external investors? Eg Does return on capital increase as stake increases?	~	~	~	~
Are people incentivized to run keepers?	*	*	~	~
It simple? Is onboarding low-friction? Where possible, do we use incentives/crypto rather than legal recourse?	~	~	*	*

3. Try New Patterns Some patterns:

- 1. Actor registry
- 2. Data registry
- 3. Actor registry + data registry
- 4. Data registry + free-as-in-beer data curation market. Curation: Pay tokens to listen.
- 5. Data registry + free data curation market. Curation: Stake tokens as belief in reputation. Auto CDN.
- 6. Actor registry + free&priced data curation market. Curation: Stake tokens as belief in reputation. Auto CDN. "Proofed Curation Market"

3. Try new patterns: evaluate on objectives & constraints

Key Question	1	2	3	4	5	6
For priced data: incentive for supplying more? Referring?	×	*	◆	ĸ	ĸ	>
For priced data: good spam prevention?	ĸ	✓	✓	✓	<	>
For free data: incentive for supplying more? Referring?	×	*	×	~	<	>
For free data: good spam prevention?	ĸ	✓	*	✓	*	 Image: A start of the start of
Does token give higher marginal value to users of the network, vs external investors? Eg Does return on capital increase as stake increases?	◆	~	~	~	<	>
Are people incentivized to run keepers?	ĸ	*	✓	✓	~	>
It simple? Is onboarding low-friction? Where possible, do we use incentives/crypto rather than legal recourse?	✓	~	*	*	~	>

Objective: maximize supply of relevant data



- Reward curating data (staking on it) + making it available
- New pattern: Proofed Curation Market

 $E(R_{ij}) \alpha \log 10(S_{ij}) * \log 10(D_j) * T * R_i$ Expected reward for user i on dataset j $S_{ij} = predicted popularity$ = user's curation market = ward e dataset = ward e dataset = ward e dataset = ward e dataset

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

$$E(R_{ij}) \alpha \log 10(S_{ij}) * \log 10(D_j) * T * R_i$$
predicted popularity
of service
proofed popularity
of service



#Token Engineering



Design of Tokenized Ecosystems From Mechanism Design to Token Engineering Synthesis: **Analysis:** Game theory Mechanism Design Practical constraints **Optimization Design** Engineering theory, practice and tools + responsibility **Token Engineering for Analysis & Synthesis**

#TokenEngineering has grown into a global movement

ot secure tokenengineering.net/community

Share on 💽 📑 📑 😭 🕺 site-name .wikidot.com Expert tip #1: Wikidot modules let you make smarter web sites

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Community

Community: Events, Related Communities, more.

TE Local Meetup Groups

(The actual meetup.com pages will typically have the n

- TE Berlin
- TE Budapest
- TE Munich
- TE Toronto
- TE London
- TE Zurich/Zug
- <u>TE Tokyo</u>
- TE NYC
- TE Amsterdam
- TE Stockholm
- TE San Francisco
- TE Vancouver
- TE Vienna
- TE St Petersburg

Wanna start your own TE meetup? Please do! :) You ca





1,448

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Ocean, Towards a Decentralized Data & Al Stack











Connective tissue: public utility network





Blockchain public utility network for the AI Commons





Have data	Have Al (Want data)	Have data	Have data	Have Al (Want data)
Have Al (Want data)	Have data	Have Al (Want data)	Have Al (Want data)	Have data
Data marketplace	DM	DM	DM	DM
DM	DM	DM	DM	DM
DM	DM	DM	DM	DM
DM	DM	DM	DM	DM

Catalyze the commons *and* profit-making, via a substrate for a thousand marketplaces to bloom.

Block rewards to incentivize data supply Add to data commons, make \$

Preserve privacy: bring AI compute to the data



XPRIZE



Framings for Ocean



"Al Compute Pipeline"



"Inter-Service Network"



"Network of Networks, for AI Stack"

Crypto perspective: Ocean is L2 on L1



Al perspective: Ocean is L1





"SABRE for data"

SABRE is metadata/substrate for airline tickets. Ocean, for data itself.



What does decentralized Al look like?

Keep Data Science Tooling, Add Crypto







Every dataset is an asset. Every Jupyter notebook is an asset (!)

ocean	Ocean	► KillTheHumans	5 Datasets	View		Login 占					
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	2	Dataset 87ertHG Facial Recognition Data	Publisher Name: VoyeurDrone Inc. Date Published: 20-06-2018	VOY	My 02 Dataset 0XhiiikA	Dataset Portfolio	4210.0CN				▲ +4.5% Current
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New powers for data scientists

- Way more data. Data commons. Enterprise data without data escapes via on-premise compute.
- **Provenance** in data & AI training. Goodbye data honeypots.
- More \$. For generating data. For cleaning, labeling, feature engineering data. For algs. For curation.





https://datascience.oceanprotocol.com

Manta Ray

ocear

Data Science powered by Ocean Protocol

The **Manta Ray** notebooks provide a guided tour of Ocean Protocol in an interactive Jupyter Notebook environment. Start using Ocean Protocol with your own pre-configured and loaded cloud instance after login with your GitHub account.

This project is in alpha! Feel free to ask questions and post bug reports in **our Gitter channel**. Notebooks are for tutorial and demonstration purpose only. Notebook instances may be periodically offline, and storage volumes will be purged.

JupyterLab Instance



Ocean Protocol 🗙 👔 Tools & Example 🗙 🗿 Quickstart - Oce 🗙	🏥 Jupyter Notebo 🗙 📔 Manta Ray — O 🗙 📄 JupyterLab	×	*	Over	view o	f the	×	+	
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QUICKSTART	<pre>git clone https://github.com/oceanprotocol/barge.git cd barge/</pre>								
Run & Try Everything	./start_ocean.shlatest								
MARKETPLACES									

Seeing the dolphin means it's working:

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Get info: oceanprotocol.com Get code: github.com/oceanprotocol Get help: gitter.im/oceanprotocol/Lobby

KEEPERS

PUBLISHERS

Run a Keeper

Set Up a Marketplace

Publish Data or Services



Conclusion

Conclusion: On Decentralizing AI Data & Services

- Goal: spread benefits of AI, by equalizing opportunities to access data
- Token design \approx EA design.
 - So, approach token design as EA design!
 - Helped a *lot* for designing Ocean. #TokenEngineering.
- Decentralized AI gives data scientists new powers
 - *Way* more data. Privacy-preserving compute.

