

# VALUING DATA AS A NEW ASSET TYPE

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based on work with Simona Abis, Juliana Begenau, Cindy Chung, Jan Eeckhout, Maryam Farboodi, Adrien Matray, Roxana Mihet, Thomas Philippon, Dhruv Singal and Venky Venkateswaran

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# BIG DATA IS RAISING BIG QUESTIONS

The most valuable firms in the world today are valued largely for their data.

Raises a whole agenda of questions, theoretical and empirical, that touch on many aspects of economics:

- ▶ Data is a new asset class. How do we price it? (finance)
- ▶ How do we measure monopoly power? What are reasonable harms to assess from data malfeasance? (law, IO)
- ▶ Is GDP measuring the value of an economy correctly? (macro)
- ▶ Does the value of data vary? Is it interest-rate sensitive? (monetary)

Our industrial-era economic tools need updating for the modern data economy.

# OUTLINE

This talk: How do we start on this agenda?

It begins with measuring the amount and value of firms' data.

- ▶ What do we mean by data?
- ▶ Data economy mechanics
  - ▶ A by-product of transactions
  - ▶ Buying and selling data
  - ▶ Depreciating data
- ▶ Measuring and valuing data: 6 approaches
- ▶ Conclude: Where next?

# WHAT DO WE MEAN BY DATA?

- ▶ Data is digitized information
- ▶ Data of interest is big data:  
Often generated by economic activity: search history, traffic patterns, purchases...
- ▶ AI / ML are prediction technologies. Data used for prediction.
- ▶ Data is distinct from tech, patents, learning-by-doing and algorithms.
  - ▶ Ideas/ technologies: procedures or concepts. Data may be an input.
  - ▶ Learning by doing: Human capital, owned by workers. Not tradeable.
  - ▶ Data is the fuel for algorithms / AI

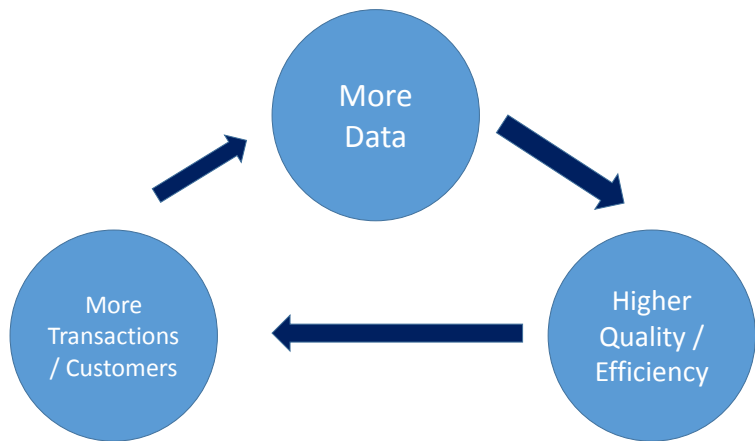
# HOW DATA IS GENERATED AND ACCUMULATED?



*"It's free, but they sell your information."*

Data price is zero. Data value is positive.

# A DATA FEEDBACK LOOP



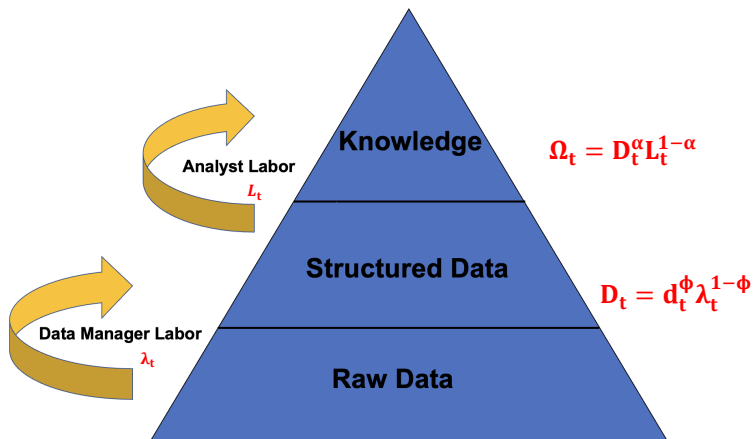
# WHAT IS HIGHER QUALITY?

## HOW DOES DATA CREATE VALUE FOR FIRMS?

- ▶ Raises current profits: Choose better products, inventory, transportation, advertise to better customers. (Tucker '22)
- ▶ Creates market power
  - ▶ Firms with more data can grow bigger, exert monopoly power.
- ▶ Reduces risk
  - ▶ Data is information. Information resolves uncertainty (risk).
  - ▶ Finance tools here are crucial.
  - ▶ This could be big: Risk compensation is 2x expected return.

# ACCUMULATING DATA: RAW DATA, STRUCTURED DATA AND KNOWLEDGE

Maybe labor is an input into useable data?





# ACCUMULATING DATA: BUYING/SELLING IT

- ▶ Data is non-rival: You can sell it and keep it.
- ▶ Data leaks: through prices, through neighbors' transactions (Acemoglu et al '22)
- ▶ Data known by others loses value (strategic substitutability).
- ▶ Firms selling data face a commitment problem: (Liu, Ma, Veldkamp '23)  
To extract monopoly rents, sell data to few clients.  
But tomorrow, data seller can sell more copies at a lower price and dilute the value of data.

# HOW DOES DATA DEPRECIATE?

- ▶ A key question for valuation.
- ▶ Ex: Data to predict an moving target:  $\theta_{t+1} = \rho\theta_t + \epsilon_{t+1}$
- ▶ Data about today's state  $\theta_t$  is less valuable tomorrow because the state  $\theta_{t+1}$  has changed.
- ▶ Similar to capital depreciation:  $k_{t+1} = (1 - \delta)k_t + i_t$ , where  $\delta = 1 - (\rho^2 + \sigma_\epsilon^2 \Omega_t)^{-1}$ .
- ▶ Data depreciates faster when it's abundant  $\Omega_t$  and the environment has volatile innovations  $\sigma_\epsilon^2$ .



# Measuring and Valuing Data

# SIX APPROACHES TO MEASURING DATA

1. Complementary inputs
2. Value functions
3. Revenue
4. Choice covariance
5. Market prices
6. Cost accounting

# MEASUREMENT APPROACH 1: COMPLEMENTARY INPUTS

- ▶ Knowledge is produced using structured data and "analyst" labor:

$$K_{it} = D_{it}^{\alpha} L_{it}^{1-\alpha}, \quad (1)$$

- ▶ New structured data is added to the existing stock of structured data with "data management" labor. Depreciates at rate  $\delta$ :

$$D_{i,t+1} = (1 - \delta)D_{it} + \lambda_{it}^{1-\phi} \quad (2)$$

- ▶ Estimate data stock from hiring and wages of each. What amount of data would make employing  $L_{it}$ ,  $\lambda_{it}$  workers at wages  $w_{Lt}$ ,  $w_{\lambda t}$  optimal? (Abis-Veldkamp '23)
- ▶ Another observable complementary input: IT capital (Bresnahan, Brynjolfsson '02)

## MEASUREMENT APPROACH 2: VALUE FUNCTION APPROACH

- ▶ The same tools macro uses to value capital work for data, with an adjusted law of accumulation.

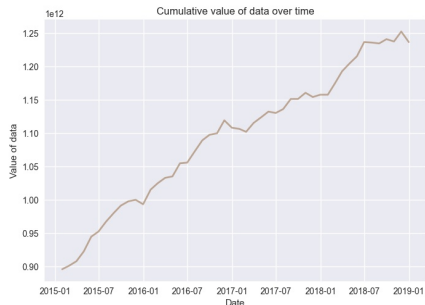
$$V(\text{data}_t) = \max_{K,L} A(\text{data}_t) K_t^\alpha L_t^{1-\alpha} - wL_t - rK_t + \beta V(\text{data}_{t+1})$$

- ▶ The state law of motion is the depreciation equation:

$$\text{data}_{t+1} = \underbrace{(\rho^2 \text{data}_t^{-1} + \sigma_\epsilon^2)^{-1}}_{\text{depreciated data}} + \underbrace{n_s \sigma_s^{-2}}_{\text{new data inflows}}$$

- ▶ Pair with theories of data inflows:
  - ▶ By-product of transactions
  - ▶ Data purchases / sales
  - ▶ Using labor to process raw data

# VALUE FUNCTION ESTIMATION RESULTS ABIS, VELDKAMP '22



**FIGURE:** Estimated Value of the Aggregate Stock of Data, used for financial analysis, in hundreds of billions of current U.S. dollars, 2015-2018.

Data value is growing for 3 reasons:

1. Firms manage more data.
2. More analysis workers make each data point more valuable.
3. Firms are becoming more productive at using AI.

## MEASUREMENT APPROACH 3: REVENUE APPROACH

- ▶ The value of data is the pdv of the revenue it generates
- ▶ How to isolate data revenue from other revenue?
- ▶ This is doable. But you need a clear idea of how data generates revenue. A model is essential to compute counter-factuals with more/less data. (Manela, Kadan '21; Davila, Parlato '21; Cong, Xie, Zhang '21)
- ▶ Problem: Data has different values to different agents (a private value asset)

Next: an example of valuing data as a private value asset, using a revenue approach.



# REVENUE VALUTION AND DATA'S PRIVATE VALUE

- ▶ Suppose data is used to purchase a portfolio of risky assets.
- ▶ Value of data in  $\mathcal{I}_{it}$  from an equilibrium with heterogeneous investors, correlated information and learning from noisy prices depends on 3 sufficient statistics:  
Expected return, variance of return and the variance of return, conditional on data.
- ▶ One can estimate these moments from return forecasting regressions.
- ▶ Finding: The same data is worth \$10 – \$1.2m, depending on the investor's wealth, investment style, price impact or trading frequency.  
(Farboodi, Singal, Veldkamp, Venkateswaran, 2022)

## MEASUREMENT APPROACH 4: CHOICE COVARIANCE

- ▶ Data allows agents to make better choices or matches.
- ▶ Better choices means actions  $q_t$  that covary with payoffs  $r_t$ .

$$E[q_t r_t] = E[q_t]E[r_t] + cov(q_t, r_t)$$

- ▶ Agents cannot achieve high covariance without information. They could have good data and no skill at analyzing it. This is a lower bound on data.
- ▶ Measure the covariance.  
Ex: Portfolio alpha or a customer click-conversion rate.  
Covariances may show up as aggregation effects: In Eeckhout-Veldkamp ('23): (firm markup - product markup) measures data

# MEASUREMENT APPROACH 5: MARKET PRICES

- ▶ Market prices for data
  - ▶ Data marketplaces: Snowflake, Datarade, etc.
  - ▶ This is the intersection of supply and demand.  
Lower bound on buyer firm's private value.
  - ▶ The value to a buyer often depends on who else knows the data.
  
- ▶ Market prices for firms
  - ▶ Market-to-book values can measure intangibles  
Crouzet-Eberly '20, Peters-Taylor '17
  
  - ▶ Already used for: Branding, patents, organizational capital, ...  
e.g., Belo, Gala, Salomao, Vittorino '21, Eisfeldt-Papanikolaou '13, '14
  
  - ▶ How to tease apart the value of data?
  
- ▶ Presumes that market participants know how to value data.

## MEASUREMENT APPROACH 6: COST ACCOUNTING

- ▶ A book value approach to valuing assets is to cumulate the sum of costly investments.  
Why not add up data costs?
- ▶ Most data is a by-product of some other economic transaction.
  - ▶ There was no explicit cost for it.
  - ▶ GAAP accounting rules only count data if it was purchased.
- ▶ An implicit cost: Data is bartered.
  - ▶ That shows up as a discount in the price of the good.
- ▶ This could work, if we can impute the data barter discount.

# CONCLUSIONS

- ▶ Data is one of the most important and highly-valued assets in the modern economy.  
Also one of the hardest to observe, measure and put a price on.
- ▶ Different approaches needed for different situations.
- ▶ Theory and measurement need to work together here.
- ▶ Next steps:
  - ▶ Quantify the data barter value.
  - ▶ Private vs. social value: Do privacy laws prevent or worsen price discrimination, by undermining competition?
  - ▶ Data accumulation and firm life-cycles. Size, scope and competition.
- ▶ Textbook coming Soon

COMING SOON

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ECONOMY

TOOLS AND APPLICATIONS

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