



THE CHANGING ECONOMICS OF KNOWLEDGE PRODUCTION
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Big Picture on Big Data



Big Picture on Big Data

- Goals: measure data and knowledge production
- Why data?
 - An endogenous source of productivity gains
 - Likely
 - Innovation policy relevant
 - Fiscal policy relevant



AV

- Measuring data (D_{it}) and knowledge production (f_{it})

$$Y_{it} = f_{it}(\{K_{it}^j\}, \{L_{it}^j\}, \{M_{it}^j\}, D_{it}, \dots)$$



AV's Magic Trick

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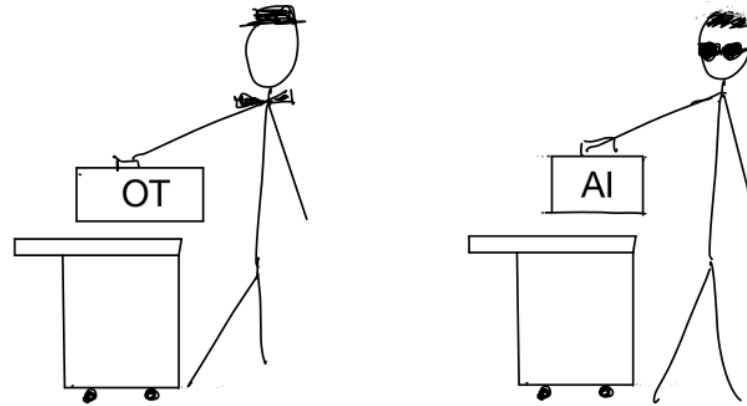


XKCD Rendition of AV



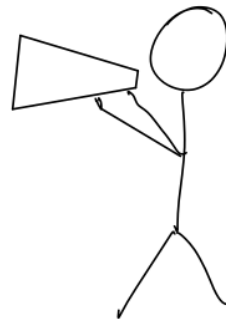
XKCD Rendition

JP Morgan



Analysts

1001
10111
000111
11100

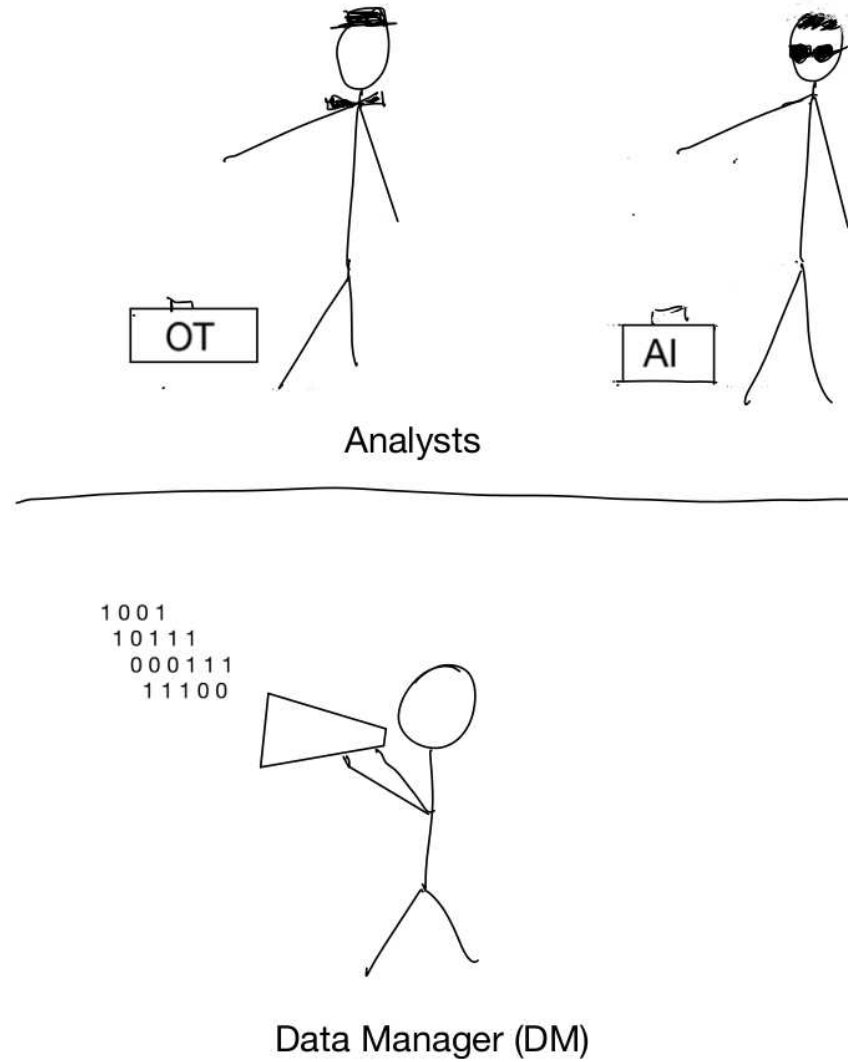


Data Manager (DM)



oops.. no desks

JP Morgan





Analysts

- Two technologies in firm i :

$$Y_{it}^{OT} = A_t^{OT} D_{it}^{\gamma} (L_{it}^{OT})^{1-\gamma}$$

$$Y_{it}^{AI} = A_t^{AI} D_{it}^{\alpha} (L_{it}^{AI})^{1-\alpha}$$

- Data manager's labor produces D_{it}
- *Note:* No other inputs or differences in TFPs



Analysts

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Analysts

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- Data manager's labor produces D_{it}
- *Claim*: $\alpha > \gamma$ suggests AI is “transformative innovation”
- What do AV do to test this?



What do AV do?

- Use Burning Glass data:
 - Skill descriptions for analysts and data managers
 - Job postings $\Rightarrow L_{it}^j, j = OT, AI, DM$
 - Wage across postings $\Rightarrow w_t^j$ (same for all $i!$)
- Solve problem of financial firm
 - Allocate analysts and managers to maximize profits



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- Solve problem of financial firm
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- How do AV identify α, γ ?



Cross-Sectional Information Not Useful

- Implication of theory:

- w_t^j = marginal product of labor_{*it*}

$$\frac{D_{it}}{L_{it}^k} = \frac{D_{jt}}{L_{jt}^k}, \quad \text{all } i, j; k = OT, AI$$

⇒ No variation in cross-section

⇒ **Cannot identify** both TFPs and shares

- If variation observed, need new theory



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- What about time dimension?



Need Variation Over Time

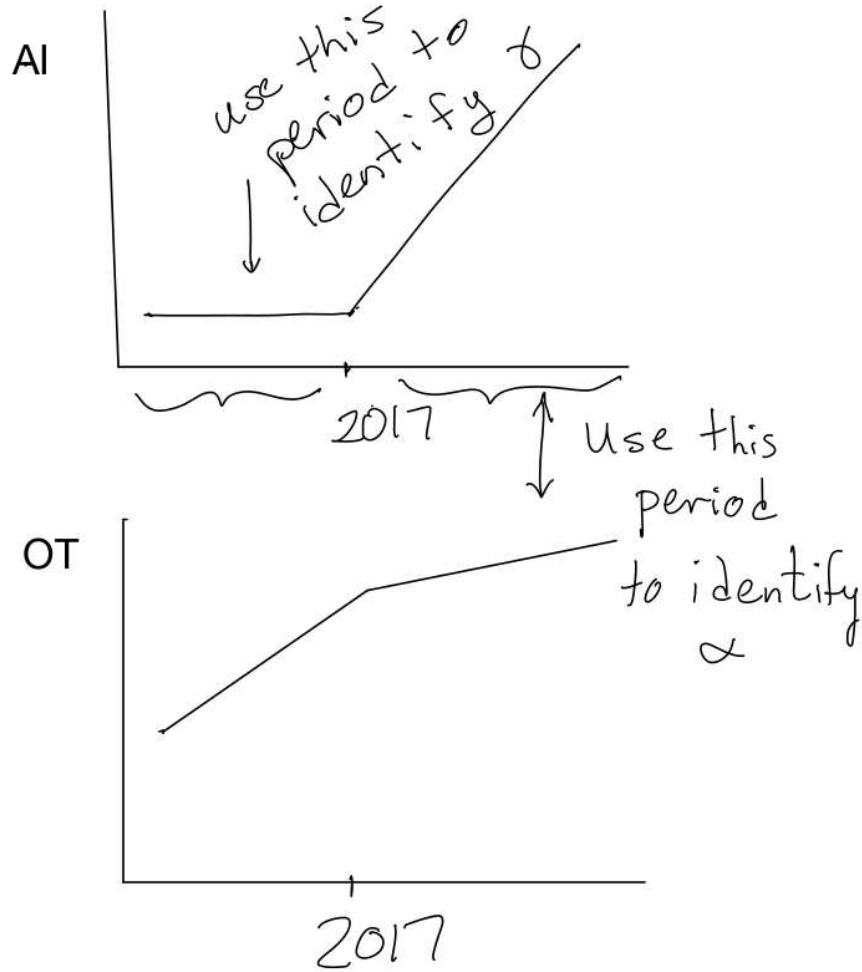
- Implication of theory:
 - Shadow price of data = marginal product of data
 - Manipulate this condition to get:

$$\Delta g(D_{it}, D_{it+1}) = \frac{\alpha}{1 - \alpha} \Delta w_t^{AI} L_{it}^{AI} + \frac{\gamma}{1 - \gamma} \Delta w_t^{OT} L_{it}^{OT}$$

- Suppose $D \propto$ wages for data managers
 - \Rightarrow Differential AI, OT earnings growth identifies α, γ



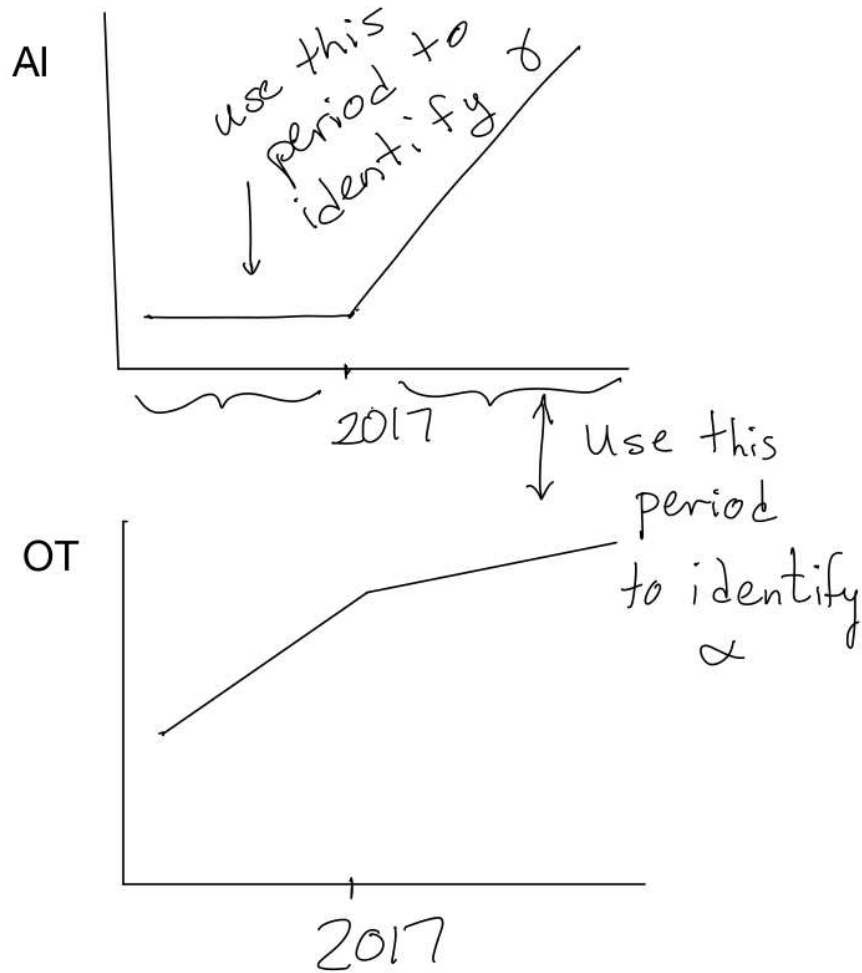
Idea Behind Identification



$$\Rightarrow \alpha > \gamma$$



Idea Behind Identification

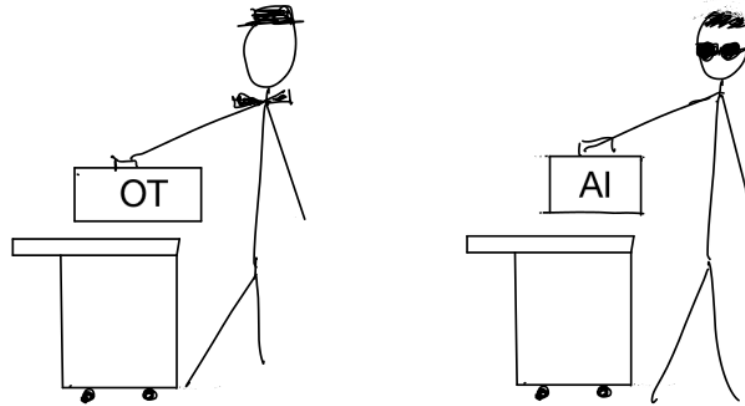


There are at least two problems here..



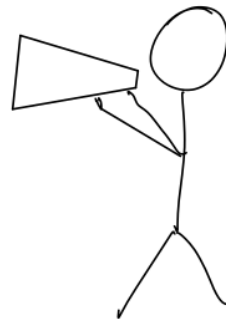
Back to XKCD

JP Morgan



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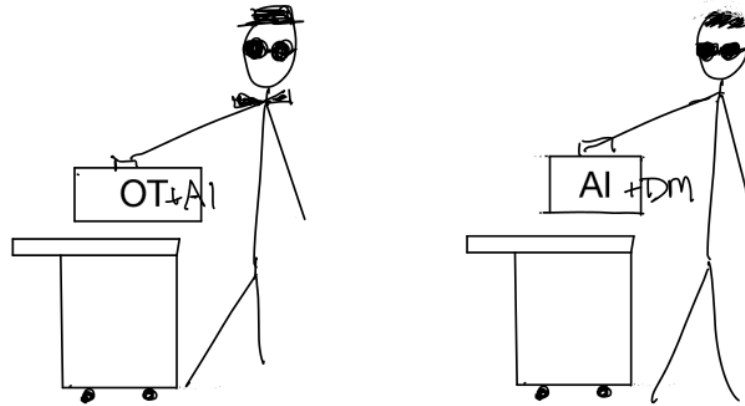


Data Manager (DM)



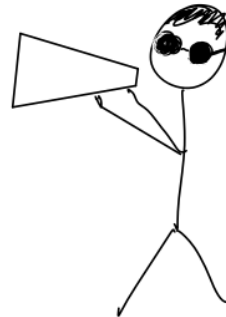
Significant Overlap of Skills

JP Morgan



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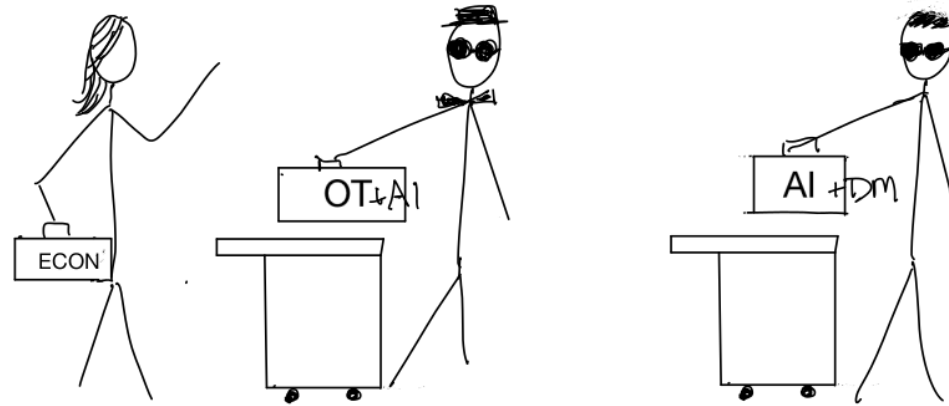


Data Manager (DM)
+AI



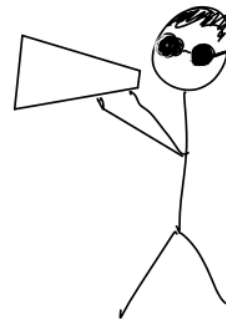
Most Analysts are Neither OT nor AI

JP Morgan



Analysts

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Data Manager (DM)
+AI



Under the Hood of Burning Glass

- Using AV's criteria for 2017, we found
 - 110+ SOC codes for OT, AI, DM
 - 92% of analysts are neither OT nor AI
- ⇒ Not obvious that distinct technologies being used



Under the Hood of Burning Glass

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- What can we learn from BLS aggregates?

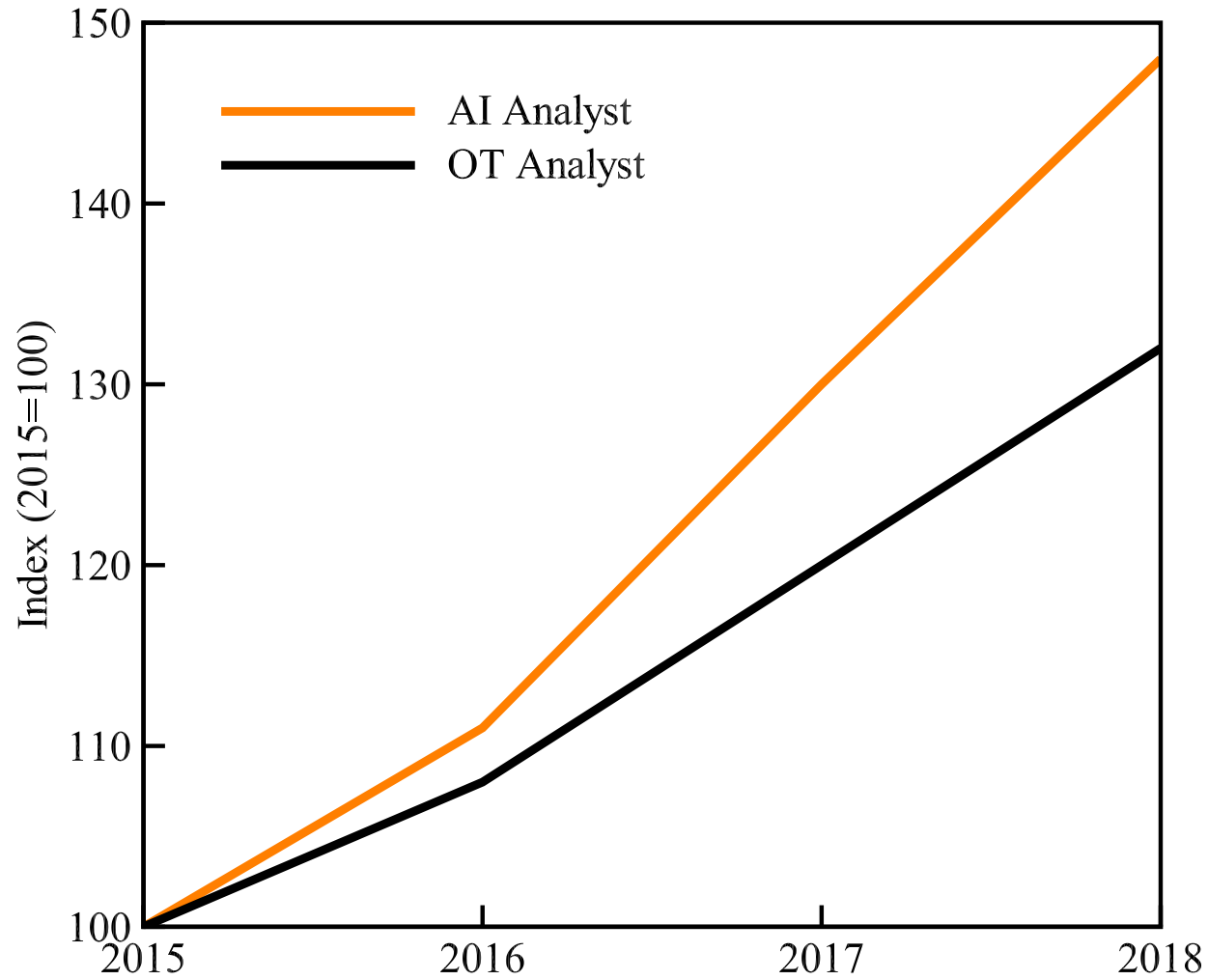


BLS Aggregates with AV Sample Weights

- Compute BLS earnings growth with AV
 - Industries
 - Occupation weights from Burning Glass
- With and without:
 - SOC 15-1199, Computer Occupations, All Other



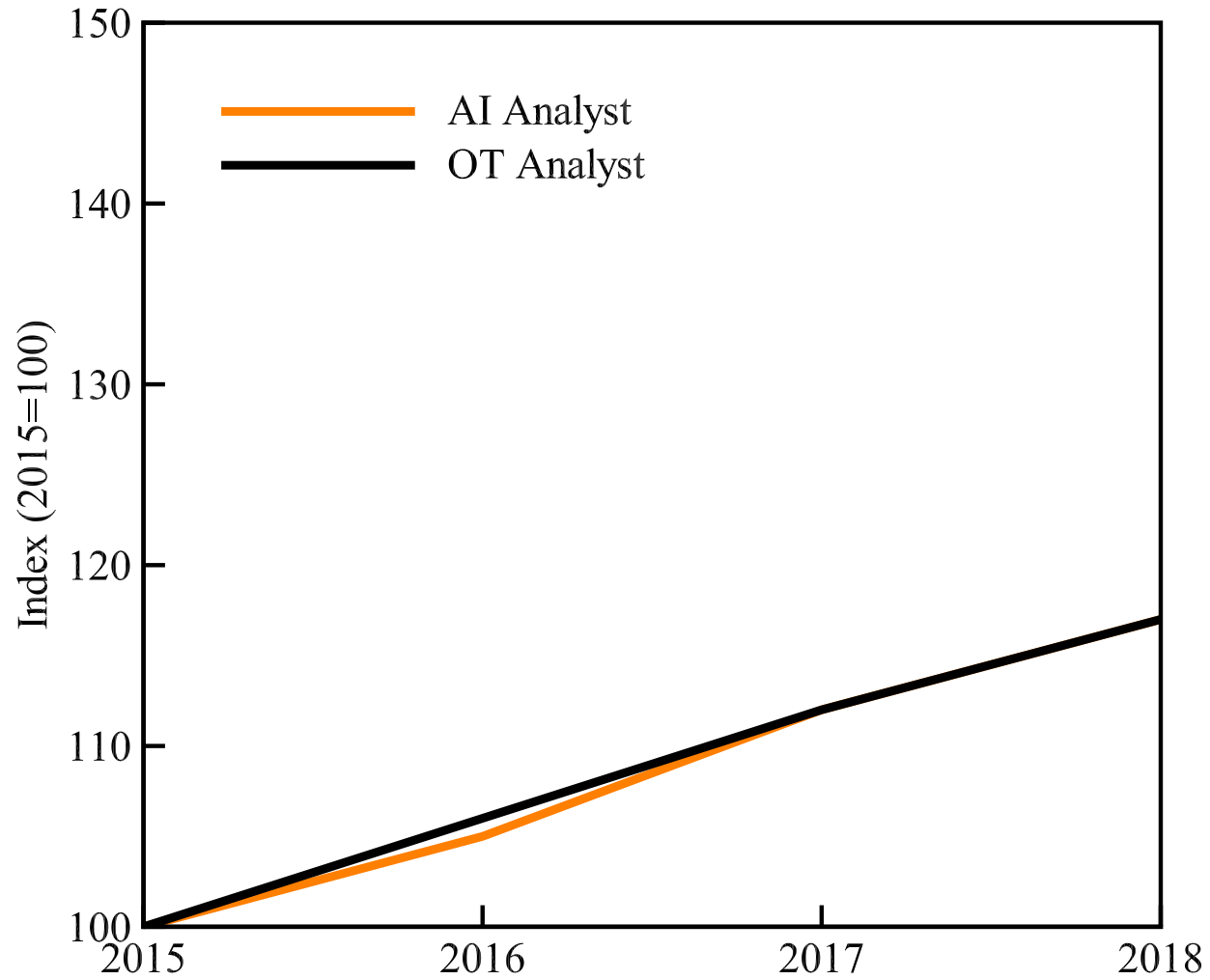
Total Earnings Growth, with 15-1199



Punchline: $\alpha > \gamma$ possible



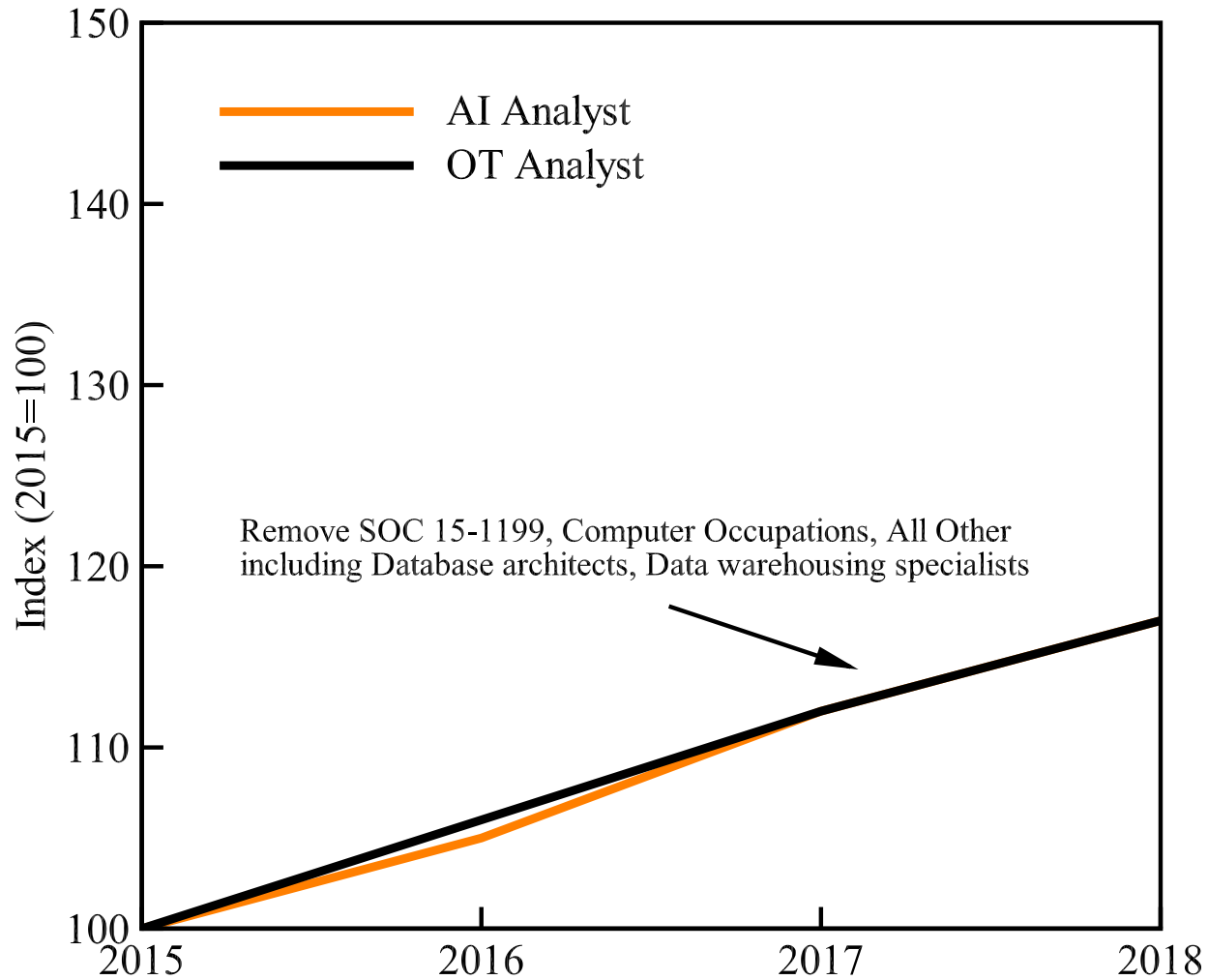
Total Earnings Growth, without 15-1199



Punchline: Results sensitive to groupings



Total Earnings Growth, without 15-1199



Punchline: AI group includes DM types



Back to Big Picture

- Good data measurement important for policy
- Need:
 - Broader scope (beyond financial services)
 - More information on production
 - Surveys like the NSF for R&D