

# Productivity Breakthrough or Not: Internet Industry and College Premium in China

UCL Explore Econ 2024  
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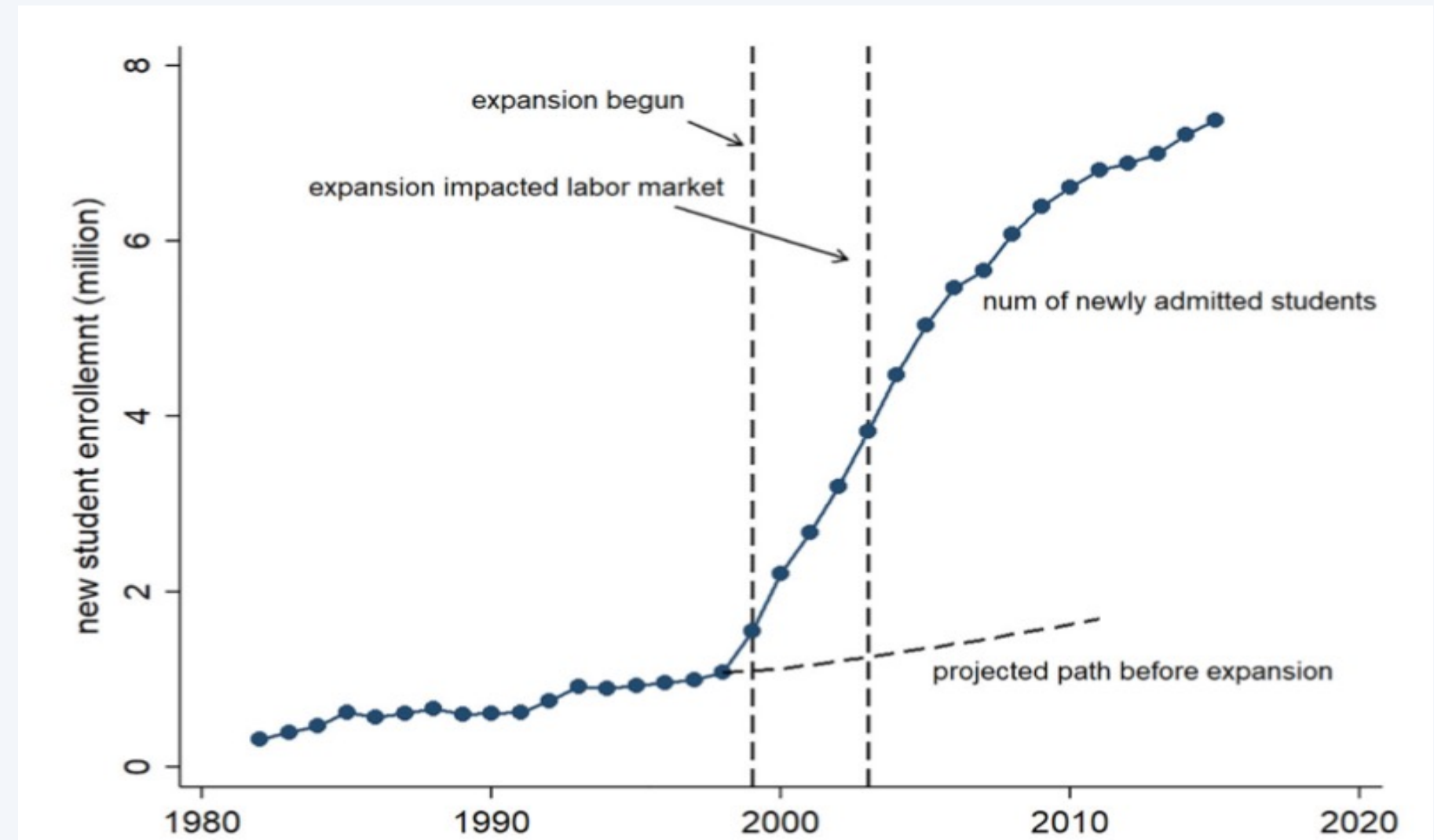




# How Has Internet Industry Impacted College Premium in China?

## The 2010s in China:

- Unusual simultaneous increase of both college premium and graduate supply
- Boom of Internet industry



## Any Connections?

- Goldin and Katz: “The Race Between Education and Technology”
- A productivity breakthrough in Internet Industry?

# Hypothesis

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**The effect of Internet industry on college premium is due to...**



**H1: Computer Application**

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**H2: Industry-Specific Productivity Breakthrough**

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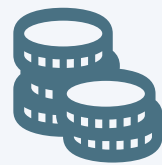
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**H2: Industry-Specific Productivity Breakthrough**



**H3: Efficiency Wage**

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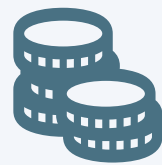
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**H1: Computer Application**



**H2: Industry-Specific Productivity Breakthrough**



**H3: Efficiency Wage**



**H4: Regional Difference**

# Methodology and Data

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## Hypothesis Verification:

- Regress log income on targeted intersection terms:  
comp for H1, indu and other\_indu for H2 and H3, prov for H4
- Control for: experience, gender, marriage, language, and employer type



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$$income = \beta_0 + \beta_1 edu + \beta_2 indu + \beta_3 other\_indu + \beta_4 edu\_indu + \beta_5 edu\_other\_indu + \beta_6 prov + X^T \sigma + \epsilon \quad (Equation 3)$$

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## Data Source

- Chinese Family Panel Study (CFPS) 2014, 2016, 2018, and 2020
- A PKU-led household survey programme, conducted every two years

# Result: Existence of College Premium and Profile for Internet Industry Worker

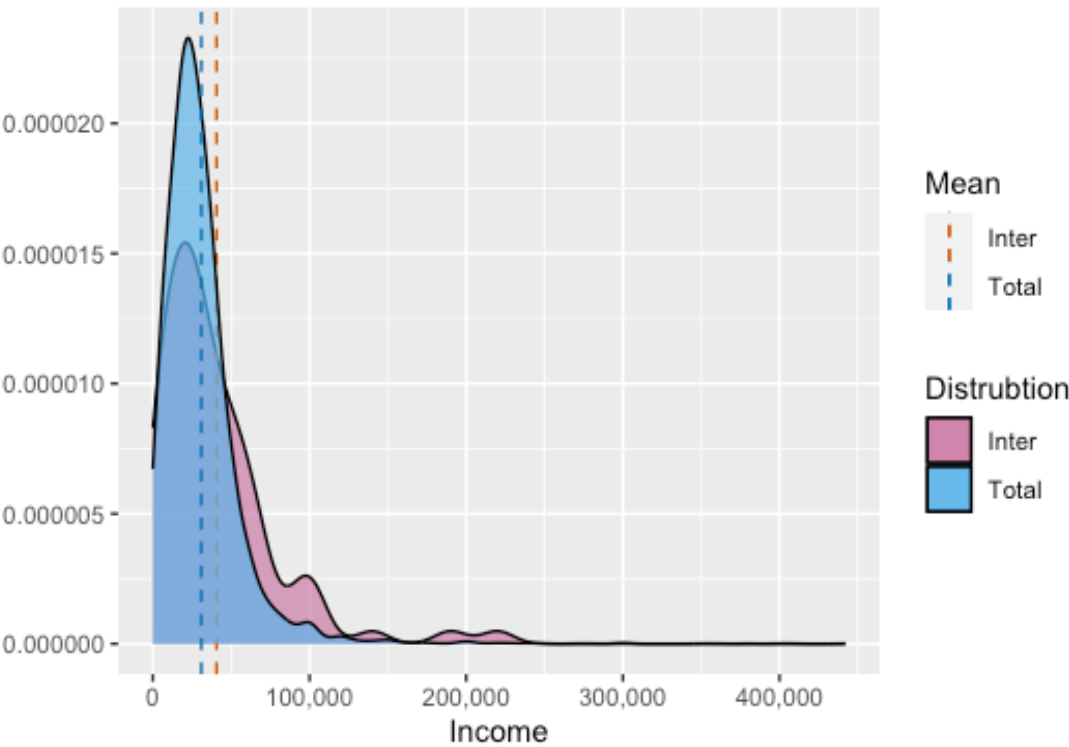
Table 4: General Regression Results

	2014	2016	2018	2020
edu	0.365*** (.091)	0.357*** (.041)	0.394*** (.023)	0.419*** (.025)
comp	0.223*** (.021)	0.098** (.032)	0.295*** (.020)	0.321*** (.021)
indu	-1.382 (.091)	0.045 (.114)	0.128* (.064)	0.083 (.062)
other_indu	0.116 (.037)	0.164** (.058)	0.117*** (.030)	0.203*** (.032)
prov	0.382*** (.020)	0.510*** (.032)	0.364*** (.018)	0.375*** (.019)
N	7394	8949	9024	8118

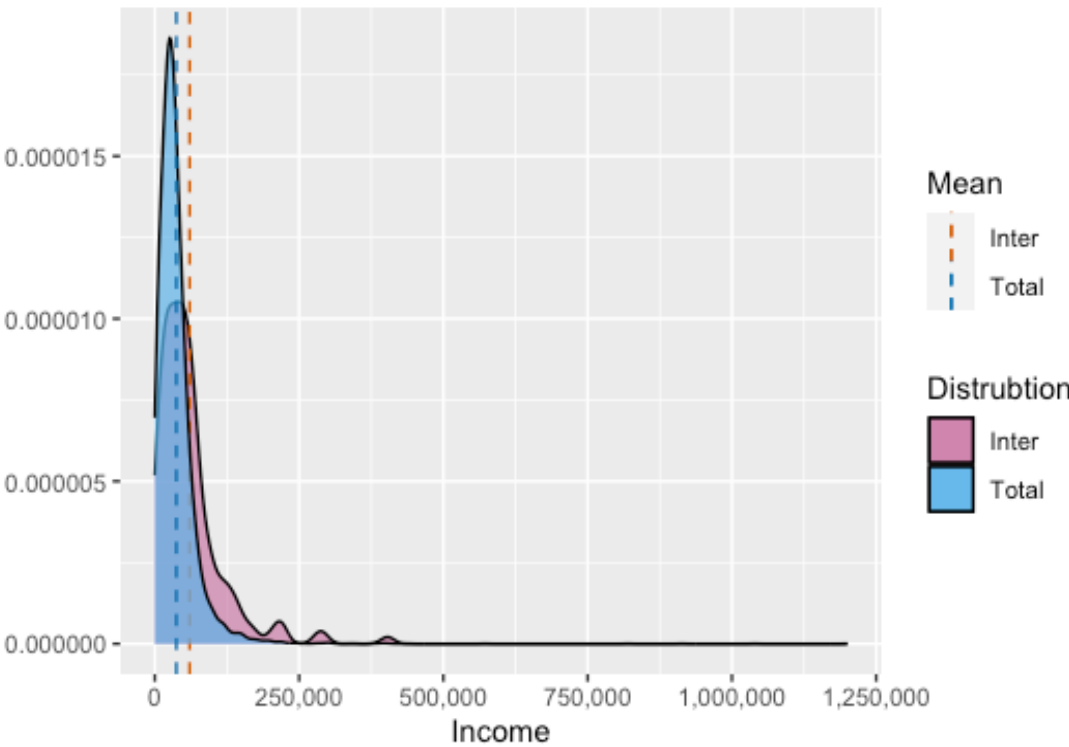
Table 3: Profile for workers in Internet industry

	income		edu		comp		prov		N	
	Total	Inter	Total	Inter	Total	Inter	Total	Inter	Total	Inter
2014	30971	40932	0.223	0.646	0.390	1.000	0.316	0.402	7394	81
2016	37387	59697	0.255	0.695	0.452	1.000	0.317	0.461	8949	164
2018	43988	74651	0.293	0.755	0.465	1.000	0.300	0.445	9024	154
2020	51832	88354	0.342	0.774	0.505	1.000	0.283	0.439	8118	163

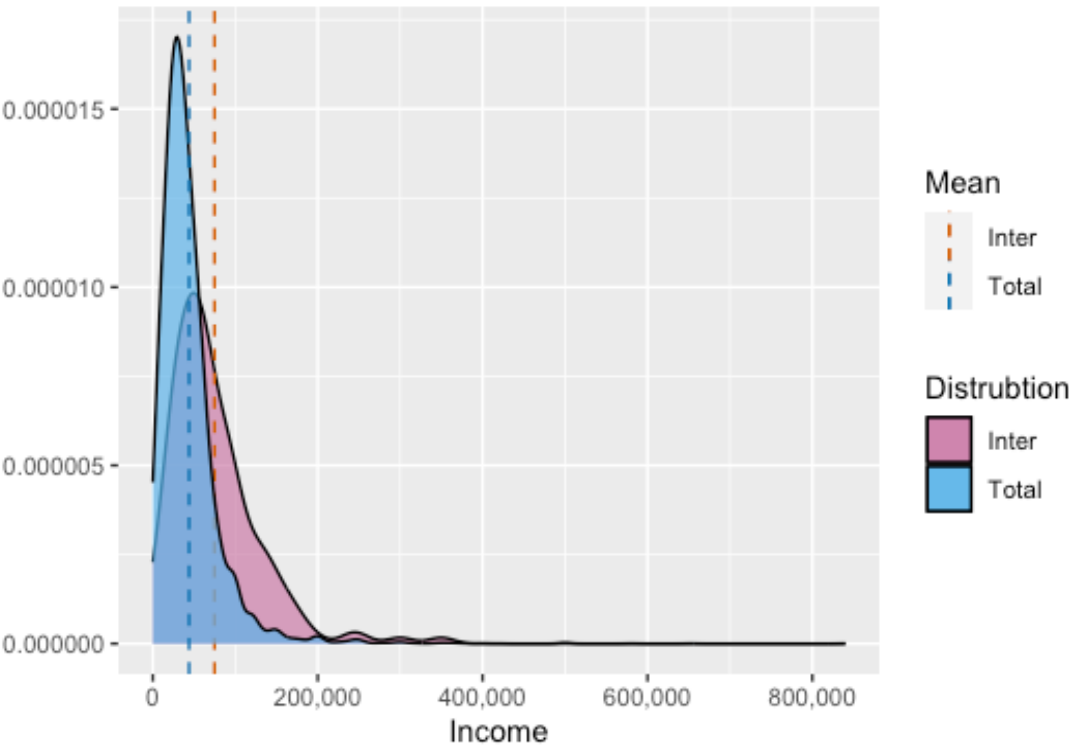
Graph 1: Internet Industry and Total Sample, 2014



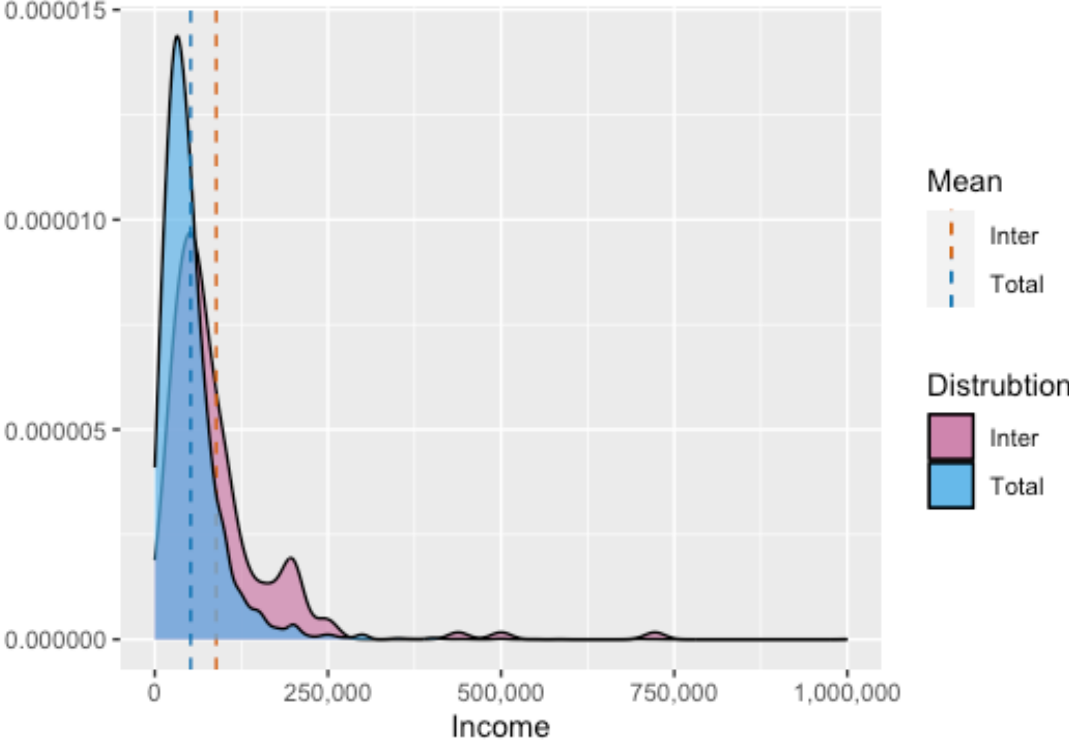
Graph 2: Internet Industry and Total Sample, 2016



Graph 3: Internet Industry and Total Sample, 2018



Graph 4: Internet Industry and Total Sample, 2020



# Result: Productivity Based Effect

Table 5: Regression Results for Equation 2

	2014	2016	2018	2020
edu	0.315*** (.051)	0.297*** (.059)	0.288*** (.044)	0.376*** (.045)
comp	0.211*** (.025)	0.075* (.036)	0.271*** (.022)	0.316*** (.024)
edu_comp	0.07 (.051)	0.114 (.059)	0.147** (.048)	0.064 (.050)
N	7394	8949	9024	8118

Note: Standard errors are presented in parenthesis.

\*  $p < 0.1$ ; \*\*  $p < 0.05$ ; \*\*\*  $p < 0.01$

Table 6: Regression Results for Equation 3

	2014	2016	2018	2020
edu	0.354*** (.029)	0.363*** (.043)	0.395*** (.024)	0.410*** (.025)
indu	-0.474** (.153)	0.282 (.206)	0.103 (.127)	-0.007 (.130)
edu_indu	0.516** (.189)	-0.338 (.245)	0.033 (.146)	0.120 (.147)
other_indu	0.115* (.045)	0.154* (.073)	0.127** (.040)	0.176*** (.043)
edu_other_indu	0.002 (.081)	0.025 (.121)	-0.022 (.062)	0.058 (.063)
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Effect from computer application (H1) in 2018

Effect from industry breakthrough (H2) in 2014



# Result: Non-Productivity Based Effect

Table 7: Regression Results for Equation 4

	2014	2016	2018	2020
edu	0.305*** (.031)	0.304*** (.047)	0.376*** (.026)	0.359*** (.027)
prov	0.335*** (.023)	0.460*** (.037)	0.340*** (.021)	0.297*** (.024)
edu_prov	0.194*** (.048)	0.178* (.073)	0.077* (.038)	0.217*** (.040)
N	7394	8949	9024	8118

Note: Standard errors are presented in parenthesis.

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- Regional difference (H4) always matters
- Indirect evidence for Internet industry effect in 2020

- No evidence for efficiency wage (H3)

# Takeaway

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## Conclusions

- Productivity based effect in 2014 and 2018
  - Industry-specific productivity growth in 2014
  - Computer-led productivity growth in 2018
- Indirect evidence for effect in 2020
- Internet industry has significant impact on college premium
- This impact mostly comes from productivity breakthrough

## Contributions

- Update understanding of college premium in China
- New prospective of impact from an emerging industry

Thanks!

