

Are They the Best and the Brightest? Analysis of Employer-Sponsored Tech Immigrants

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- Industry wants more H-1B work visas, and fast-track green cards for STEM foreign students.

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- How do rates of top foreign talent vary from employer to employer?
- What are rates of top foreign talent among the main nationalities, i.e. Chinese and Indian?

Previous Work

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 - Many Indian-immigrant firms are in the outsourcing business.

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- GAO even noted role of loopholes:

... [employers] hired H-1B workers in part because these workers would often accept lower salaries...however, these employers said they never paid H-1B workers less than the required wage.

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- Artificially “raise” H-1B salaries by factor 1.2.

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- Among those who (ever) came to U.S. as foreign students in CS/EE and were working in CS/EE as of 2003, 23.2% were Chinese and 27.2% were Indian.

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year	China	India
2005	0.134	0.444
2006	0.103	0.501
2007	0.097	0.515
2008	0.080	0.569
2009	0.075	0.590

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- So, only (median) WR values higher than, say 1.25, indicate a firm is hiring mainly the “best and brightest” foreign workers.

Overall PERM Results

Median WR values (SE=sw. engineers: EE=elec. engineers):

group	med. WR
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EE	1.00
Chinese SE	1.02
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- no evidence that the foreign students are outperforming the domestic ones

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India	13.8% \pm 9.1%	19.0% \pm 7.0%

- Chinese award rate much lower than average
- Indian rate could be about average

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- Governments of China, Japan, S. Korea and Taiwan have all tried to remedy this.

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- The tech industry is famously meritocratic for engineering (not managerial) workers. If you produce, you are rewarded. English is not a major issue.
- Logistic regression analysis on the PUMS census data shows that among immigrant Chinese, English skill has no impact on the probability of earning a high-level salary ($> \$150K$).

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- Employers still very picky in hiring, again counterindicating a shortage.

Internal Brain Drain

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- “Innovation” is the buzzword *de jour*, and it is U.S.’ only comparative advantage. Yet the system is wasting that advantage.

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- Should transfer much of the EB-3 quota to EB-2.