

Discussion:

Cross-country differences in the size of venture capital financing rounds

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Summary

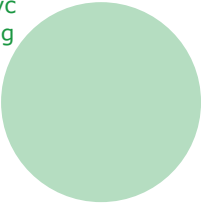
- **Outcome:** $y :=$ Size of venture financing rounds
- **Analysis:** $y = f(\hat{X})$
- **Model:** Predict \hat{X} with ML
- **Results:**
 1. Firm- and country characteristics $\rightarrow y$
 2. Degree of development of venture capital industry $\nrightarrow y$

Comments: Main Result

Degree of development of venture capital industry $\neq X$

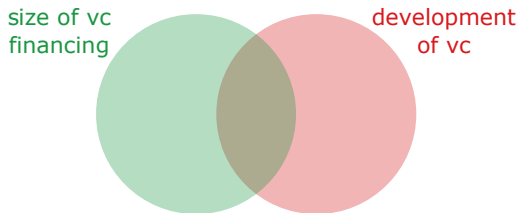
Lack of identifying variation \neq Lack of causal effect

size of vc
financing



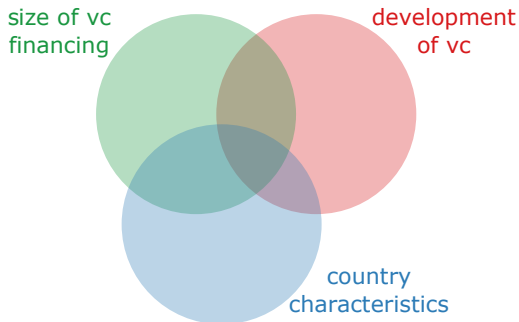
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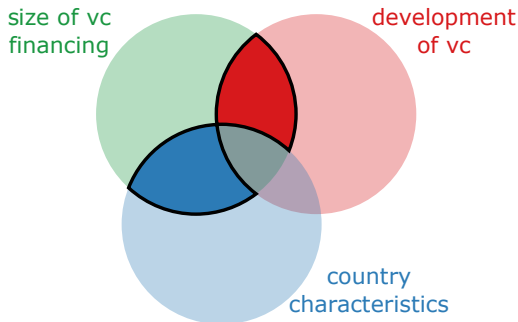
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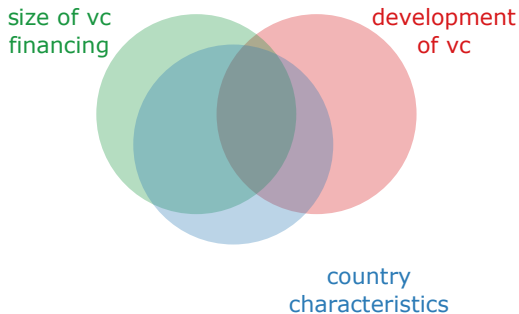
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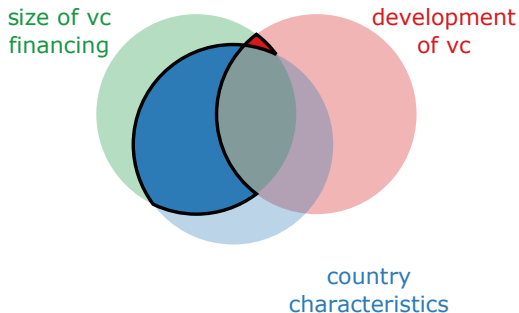
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Comments: Main Result

Degree of development of venture capital industry $\notin X$
Lack of identifying variation \neq Lack of causal effect

Possible solution

- Redundancy argument (already in the paper)
 - You show a high correlation
(R^2 of 50% compared to R^2 of 10% in main analysis)
 - Disentangle relationship between variables by moving away from lower dimension (correlation between individual components)
- Move away from agnostic approach
 - More careful selection of control variables
 - Exploit variation over time (if existent)

Comments: Inference

Use of cluster-robust standard errors

- Standard errors **incorrect** for two-step procedure:
 1. Generate \hat{X}
 2. Estimate $y = f(\hat{X})$ with cluster-robust standard errors

Comments: Inference

Use of cluster-robust standard errors

- Standard errors **incorrect** for two-step procedure:
 1. Generate \hat{X}
 2. Estimate $y = f(\hat{X})$ with cluster-robust standard errors
- Ignores sampling variance in the first-stage estimates
- Similar to 2SLS in IV-estimation, we have to correct the standard errors
 - Angrist & Pischke 'Mostly Harmless Econometrics' Ch. 4
 - Wagner & Athey (2019) 'Estimation and Inference of Heterogeneous Treatment Effects using Random Forests'

Standard errors are **incorrect for two-step procedure**

Possible solution

- Asymptotic results likely not available
- Use block-bootstrap for entire two-step procedure

Comments: Machine Learning

Use of specific ML procedure is not properly motivated

Alternative approaches for dimensionality reduction

- PCA
- Neural Networks (Autoencoder)
- Selection by domain experts

Possible solution

- Show results with alternative approaches
- Evaluate benefits of ML procedure (gain in efficiency?)
- Compare out-of-sample predictive accuracy

Comments: Machine Learning

Interpretability

- Interpretation of marginal effects of \hat{X} is very difficult
- Agnostic ML approach bears many risks

Example: The World Bank's Starting a Business Scores of Germany and Sierra Leone are very similar

Possible solution

- ML procedure for robustness check
- Main analysis:
 - Panel model with fixed effects; show within and between R^2 , or
 - Careful variable selection and modelling choices (e.g., interactions with OECD dummy, ...)

Minor Points

- Low in-sample fit ($R^2 \approx 10\%$)

Ideas on the source of unexplained variation?

- Null effect as main result

Convince reader that null is precisely estimated (e.g., by showing confidence interval)

- Do not interpret size of insignificant coefficients
- “Boosted trees and stacked generalization allow us to construct variables that summarize all the information. . .”

Dimensionality reduction always implies loss of information