

# Robustness Checks in Structural Analysis

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# Outline

- 1 Summary
- 2 Issue
- 3 Robustness
- 4 Conclusion

# I am going to do three things!

- ▶ Explain what is going on
- ▶ Identify an issue and offer a solution
- ▶ Talk about the importance of the topic

Find an easy way to do robustness checks regarding moment selection in structural estimation.

- ▶ Why does this need to be easier?
- ▶ Structural estimation is sssllllloooowwww.
- ▶ It is hard to do robustness if it takes one week to do an estimation.

## Why is structural estimation slow

- ▶ Estimate some parameters,  $\theta$ , of an **economic** model by minimizing

$$\left[ \begin{array}{c} \text{data} \\ \text{moments} \end{array} - \begin{array}{c} \text{simulated} \\ \text{moments} \end{array}(\theta) \right]' \hat{W} \left[ \begin{array}{c} \text{data} \\ \text{moments} \end{array} - \begin{array}{c} \text{simulated} \\ \text{moments} \end{array}(\theta) \right]$$

weight matrix

- ▶ **simulated moments**  $(\theta)$  can be very nonlinear
- ▶ no closed form

► Minimizing

$$\left[ \begin{array}{c} \text{data} \\ \text{moments} \end{array} - \begin{array}{c} \text{simulated} \\ \text{moments} \end{array} (\boldsymbol{\theta}) \right]' \hat{\mathbf{W}} \left[ \begin{array}{c} \text{data} \\ \text{moments} \end{array} - \begin{array}{c} \text{simulated} \\ \text{moments} \end{array} (\boldsymbol{\theta}) \right]$$

is like trying to walk down:



► blindfolded

## Walking down the mountain takes time

- ▶ No closed form
- ▶ With each step you have to
  - ▶ solve the model
  - ▶ simulate data
  - ▶ recalculate the bumpy thing to be minimized
- ▶ It can take 50,000 or more steps.
- ▶ Estimating once is usually fine. Many times? Not so much.

## Illustrate their alternative with a very simple investment model

- ▶ A firm maximizes the expected present value of shareholder distributions
- ▶ Distributions are profits minus investment
- ▶ Profits come from an exogenous **shock** and a **decreasing returns function of capital**
- ▶ The firm invests in capital

$$\max_{k_{t+j}, j=0, \dots, \infty} \mathbb{E}_0 \sum_{j=0}^{\infty} \frac{1}{(1+r)^t} \left( z_{t+j} k_{t+j}^{\alpha} - \left( k_{t+j+1} - (1-\delta)k_{t+j} \right) \right)$$

$$\max_{k_{t+j}, j=0, \dots, \infty} \mathbb{E}_0 \sum_{j=0}^{\infty} \frac{1}{(1+r)^t} \left( \text{profits} - \text{investment} \right)$$



# The solution is “marginal product of capital equals user cost!”

- ▶ It is that simple

$$E_t (\alpha z_{t+1} k_{t+1}^{\alpha-1}) = r + \delta$$

$$E_t (\text{MPK}_{t+1}) = \text{interest rate} + \text{depreciation rate}$$

$$= \text{user cost}$$

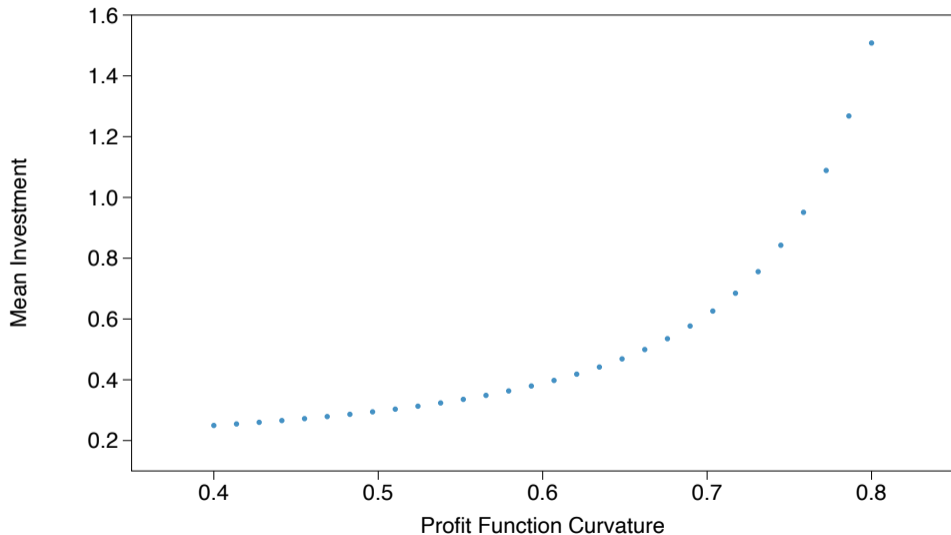
# I want to estimate profit function curvature ( $\alpha$ )

- ▶ Let's say I know  $\delta$ ,  $r$ , and the driving process for  $z_t$ .
- ▶ And I estimate  $\alpha$
- ▶ But ... I am unsure if I picked the right moments.

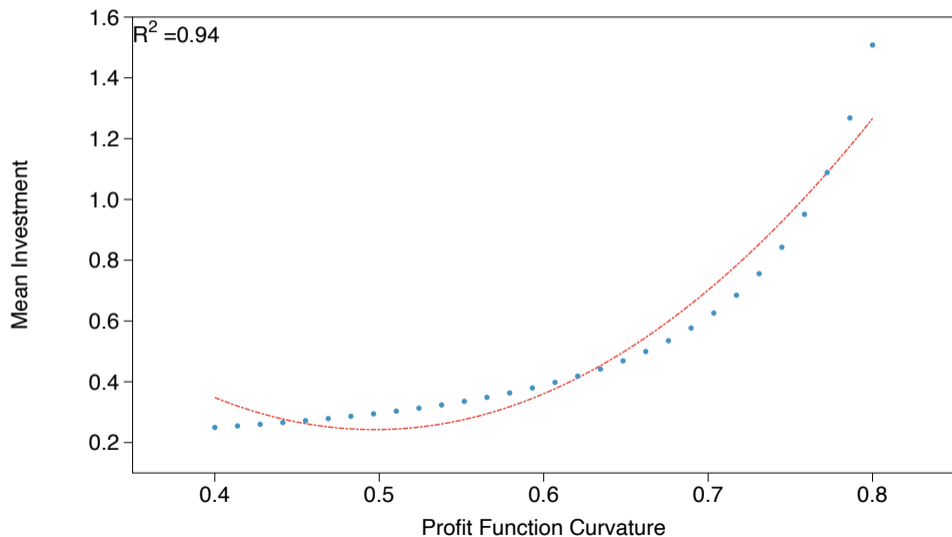
# I could just reestimate using all possible combinations of available moments

- ▶ Or I could use their method
  - ▶ Start by choosing carefully selected, random-ish values for  $\alpha$  (Sobol' or Halton sequences)
  - ▶ For each I solve and simulate the model and calculate moments.

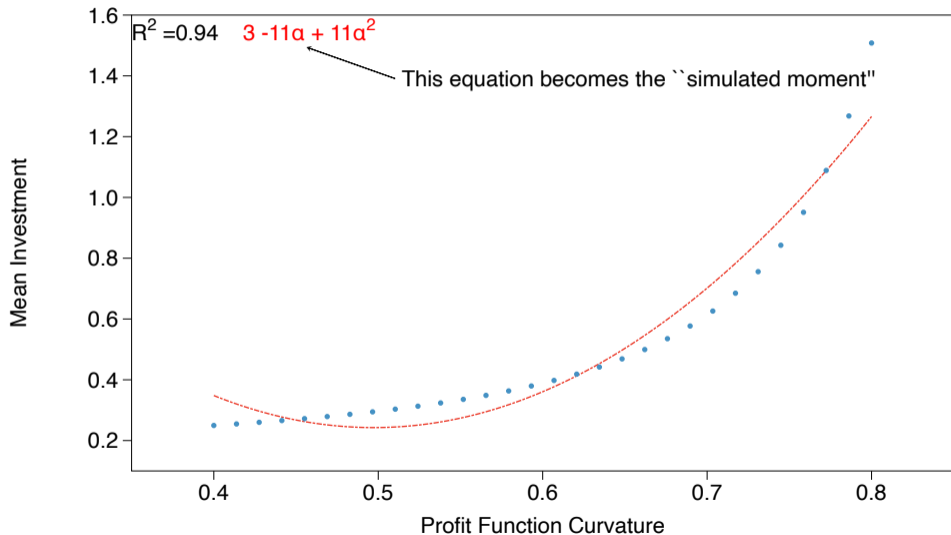
So I make the moments for each value of  $\alpha$



## And I fit the moment to the parameter



## And I fit the moment to the parameter



## Do the same thing for a bunch of other moments

- ▶ Before, computing the objective function was a process:
- ▶ Now!

$$\left[ \begin{array}{c} \text{data} \\ \text{moments} \end{array} - \begin{array}{c} \text{approximate} \\ \text{moments} \end{array}(\boldsymbol{\theta}) \right]' \hat{\mathbf{W}} \left[ \begin{array}{c} \text{data} \\ \text{moments} \end{array} - \begin{array}{c} \text{approximate} \\ \text{moments} \end{array}(\boldsymbol{\theta}) \right]$$

- ▶ has a closed form
- ▶ easy to try all possible moment vectors

# Outline

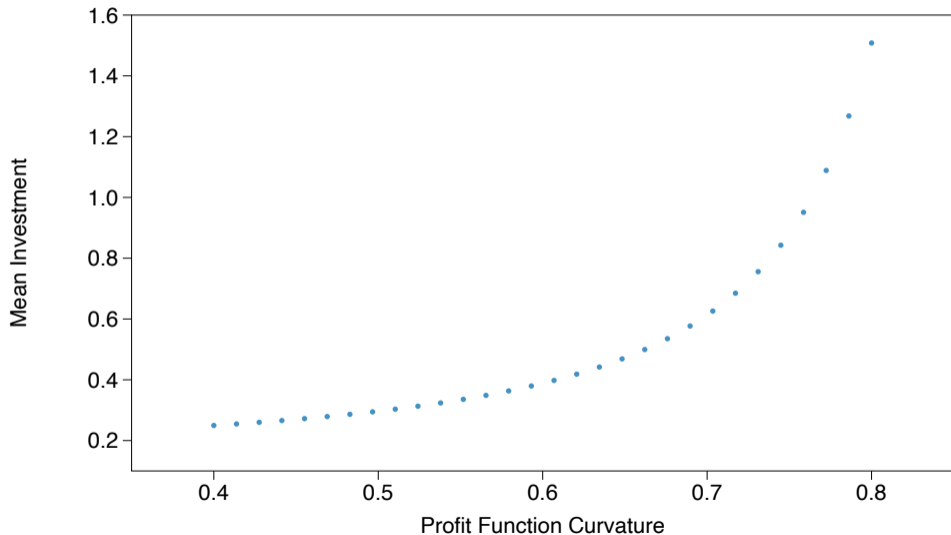
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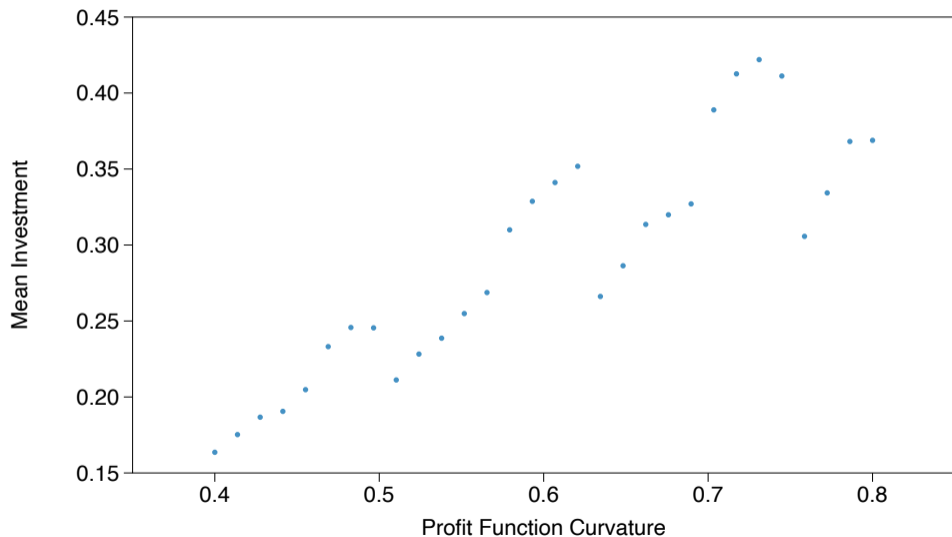
## Especially in corporate finance, these plots are often bumpy

- ▶ Frictions and nonconvexities are the bread and butter of corporate finance
- ▶ I am going to add some nonconvexities to the model
  - ▶ Non-negative profit constraint (no equity issuance)
  - ▶ Fixed cost of adjusting the capital stock

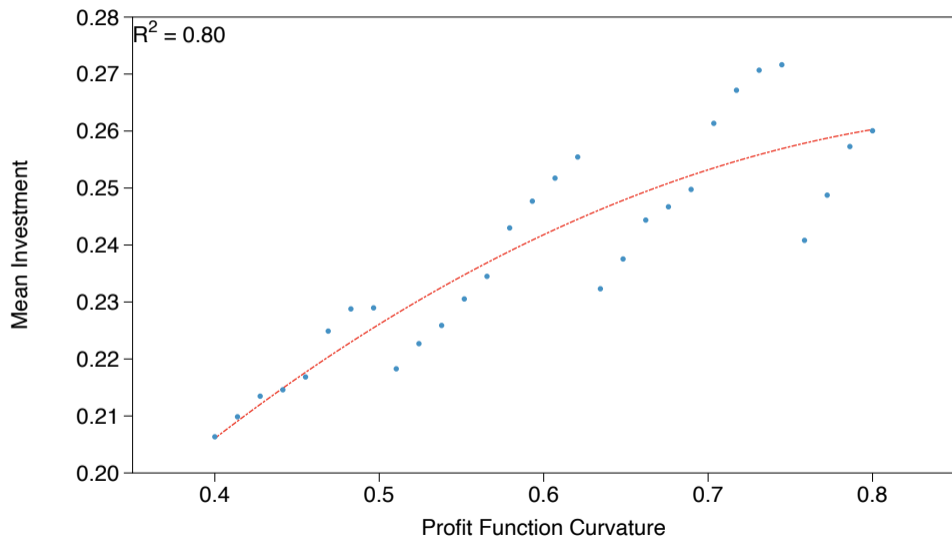
## Remember how the plot used to be smooth?



## It's not anymore with the nonconvexities



## And the fit is not as good



## But is this serious? I answer with Monte Carlos

- ▶ How often does approximate SMM recover the true value?
- ▶ Two Monte Carlos:
  - ▶ Using the actual bumpy, non-closed-form moment functions
  - ▶ Using the smooth, closed-form approximate moment functions
- ▶ Just with one specific choice of moments
  - ▶ Mean investment
  - ▶ Variance of investment

# Monte Carlo Design

- ▶ Solve model at  $\alpha = 0.7$ .
- ▶ Simulate 1000 small data sets ( $n = 500$ )
- ▶ Estimate the curvature ( $\alpha$ ) parameter using actual and approximate SMM:

## The results are not great for approximate SMM

- ▶ Approximate SMM is screamingly fast.
- ▶ And very accurate at hitting the wrong spot.
- ▶ All results are expressed as a fraction of the true value

	Actual SMM	Approx SMM Quadratic	Approx SMM Cubic
Mean Bias	0.009	0.279	0.163
RMSE	0.040	0.294	0.163

# The authors should do a bunch of Monte Carlos

- ▶ Look at models with **serious nonconvexities**
- ▶ See how good an approximation you need to **make the estimator work.**
- ▶ You can use a **laptop.**



## The authors do look at the accuracy of the approximations

- ▶ But not directly at the implications for estimation.
- ▶ This would improve the paper a great deal.

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## Is moment choice the right dimension of robustness?

- ▶ Referees never ask me to check robustness with respect to moment choice.
- ▶ Models should **not** be “robust” to the choice of moments.
- ▶ They are models, not reality.
- ▶ They **will** break.

# The set of moments should be tailored to the economic question

- ▶ If your question is about **A** ...
- ▶ and you can match moments related to **A** ...
- ▶ and you are transparent about the model's inevitable limitations,
- ▶ that's as good as it can get

# A more important dimension of robustness is the **model**

- ▶ All referees ask me about this question.
- ▶ Most structural robustness sections are about this question.
- ▶ It's the right question to ask.

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## The method in this paper might end up being quite useful

- ▶ I want to know how much **estimation** error it introduces
- ▶ I want to know if you can reduce this error with better approximations
- ▶ I do not think the technique will be useful for “robustness.”
- ▶ The technique might be useful to identify where models break.
- ▶ We can **learn** something from where the models break down.