

# Homotopy

# Philipp Müller

Gregor Reich

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## Motivational Example

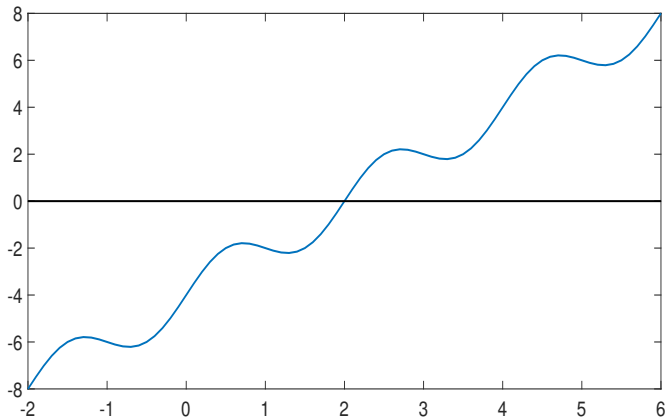
- Consider the problem of finding the root of

$$F : \mathbb{R} \rightarrow \mathbb{R}, \quad F(x) := 2x - 4 + \sin(\pi x) = 0.$$

- Nonlinear equations are
  - **omnipresent** in economics, and
  - can be **hard to solve**.
- They implicitly define, e.g., the equilibria in dynamic models and competitive general equilibria.

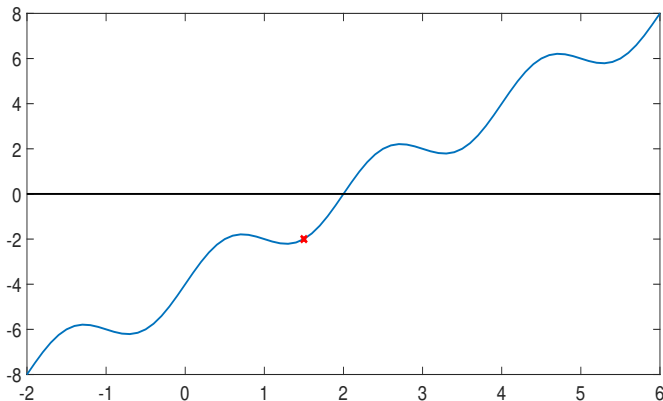
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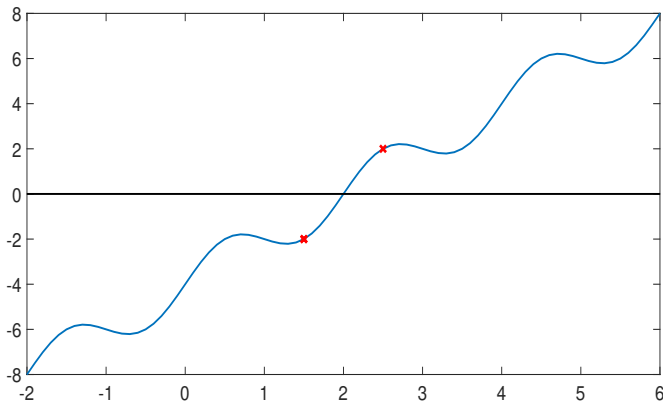
## Motivational Example: Apply Matlab's **fsolve**

$$F(x) := 2x - 4 + \sin(\pi x) = 0, \quad x_0 = 1.5$$



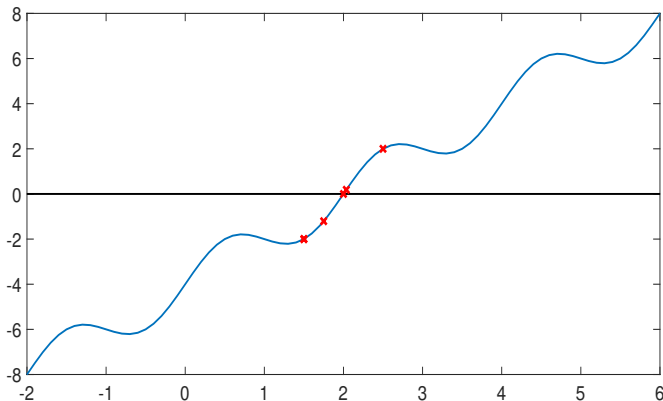
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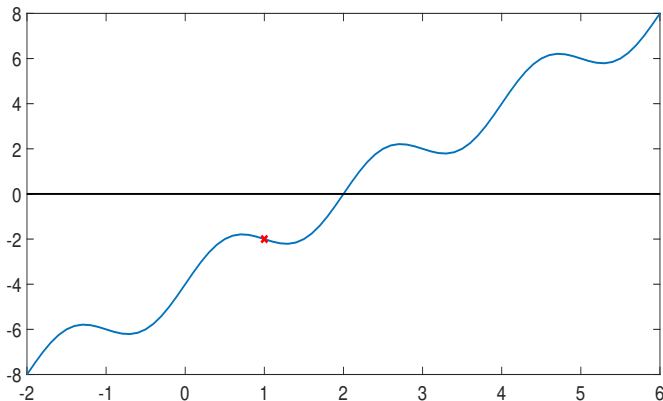
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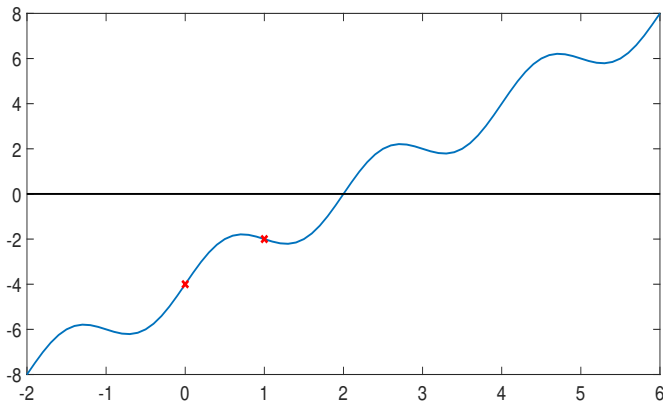
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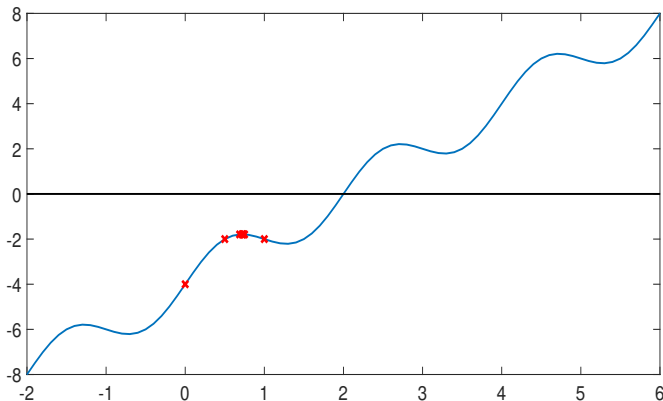
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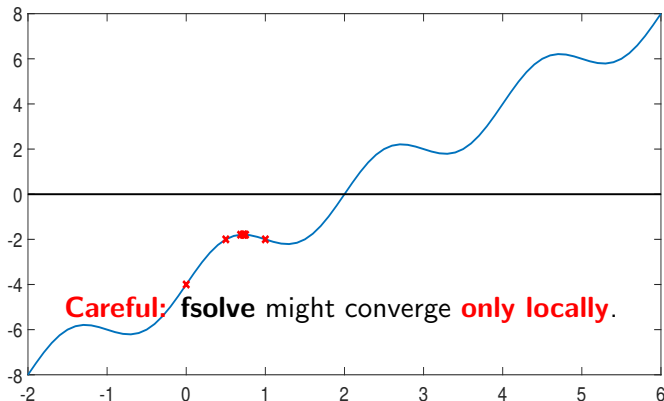
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# Simple Continuation Method: Deformation

**Deformation** Start from a *simple* function  $g$  and deform  $g$  into our target function  $F$ .

Define the homotopy map  $H(x, \lambda_i)$  as

$$H : \mathbb{R}^N \times [0, 1] \rightarrow \mathbb{R}^N, \quad H(x, \lambda) = (1 - \lambda)g(x) + \lambda F(x),$$

with the homotopy parameter  $\lambda$ .

Note: It starts at  $H(x, \lambda = 0) = g(x)$  and ends at  $H(x, \lambda = 1) = F(x)$ .

# Simple Continuation Method: Algorithm

**Objective** Find the root of  $F(x) = 0$ .

**Initialize** Define the homotopy map  $H(x, \lambda) := (1 - \lambda)g(x) + \lambda F(x)$ .

**Step 1** Start at  $\lambda = 0$  and solve  $H(x, \lambda = 0) = g(x) = 0$  for an **arbitrary** starting point  $x_0$ .

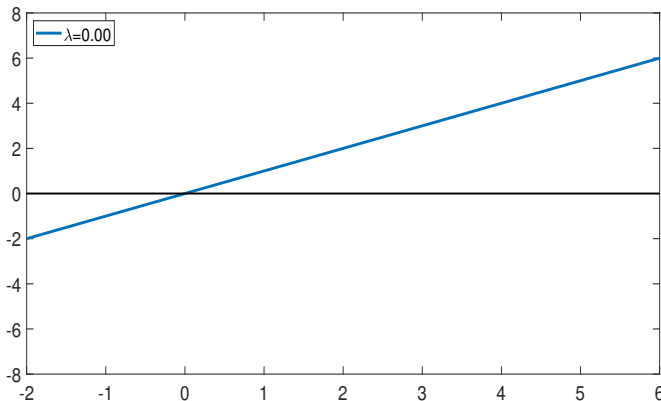
**Step 2** Increase  $\lambda$  step-wise and solve  $H(x_i, \lambda_i) = 0$  **for each**  $\lambda_i$ .  
Use the solution from the **previous step**  $i - 1$  as start point.

**Result** For  $\lambda = 1$ , we found the solution  $\bar{x}$  solving  $H(\bar{x}, 1) = 0$  and

$$F(\bar{x}) = 0$$

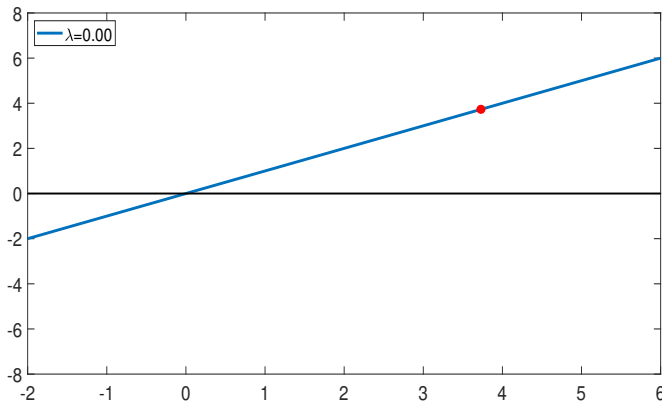
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$$H(x, \lambda) := (1 - \lambda)x + \lambda(2x - 4 + \sin(\pi x)) = 0$$



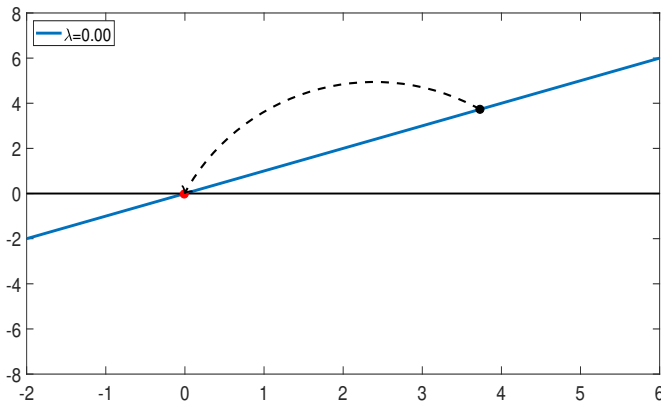
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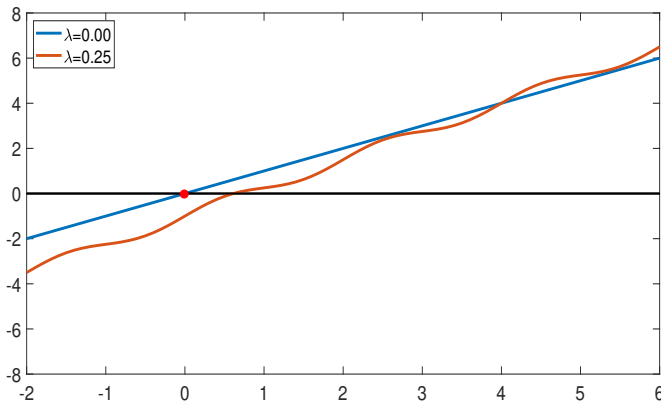
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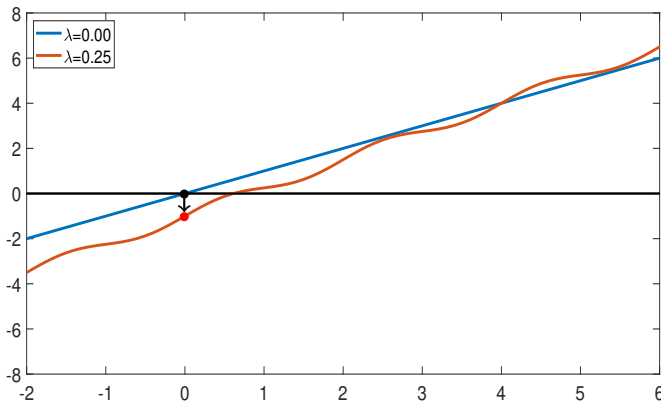
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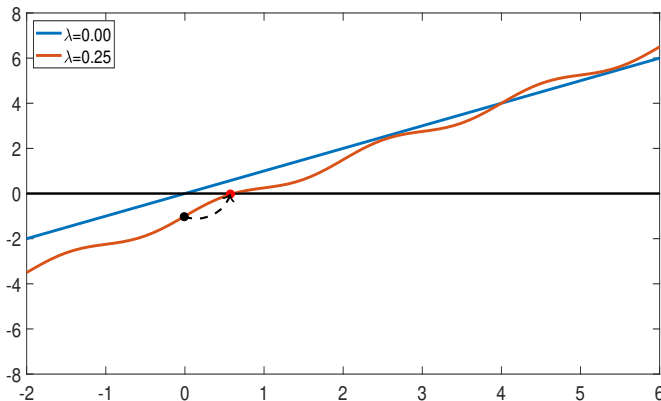
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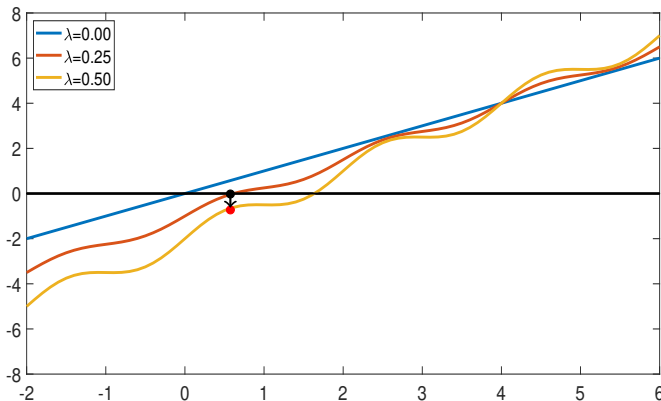
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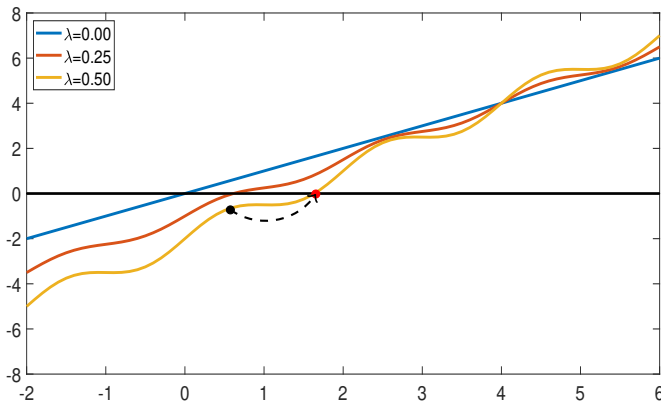
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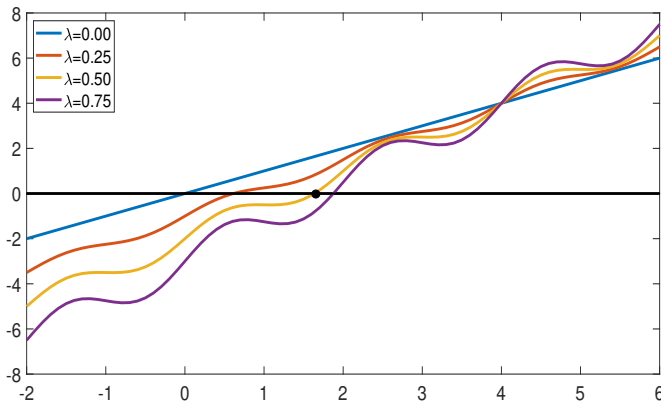
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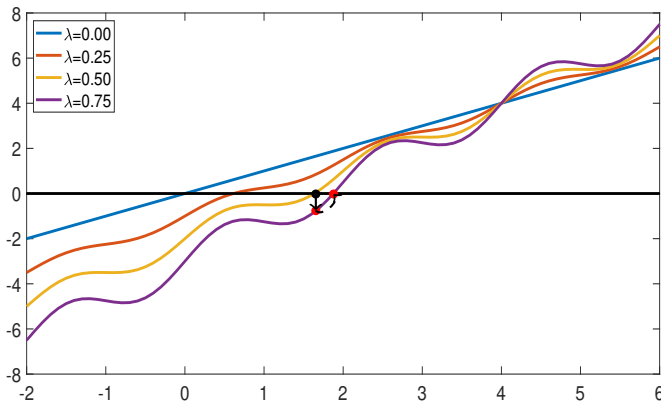
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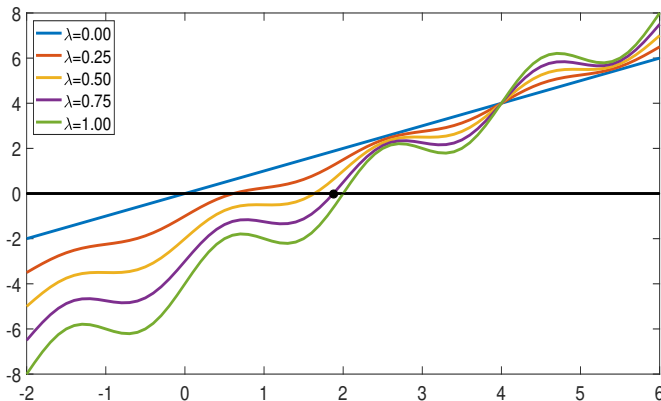
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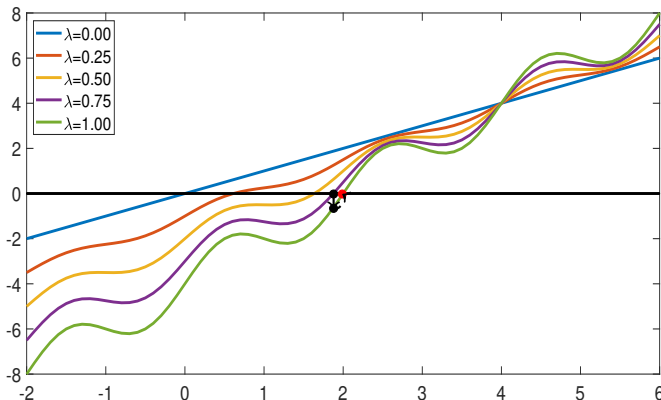
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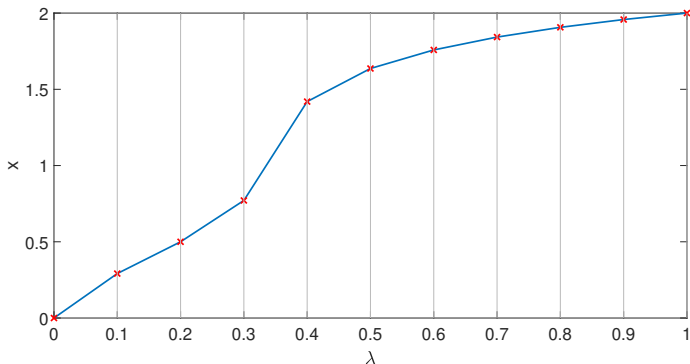
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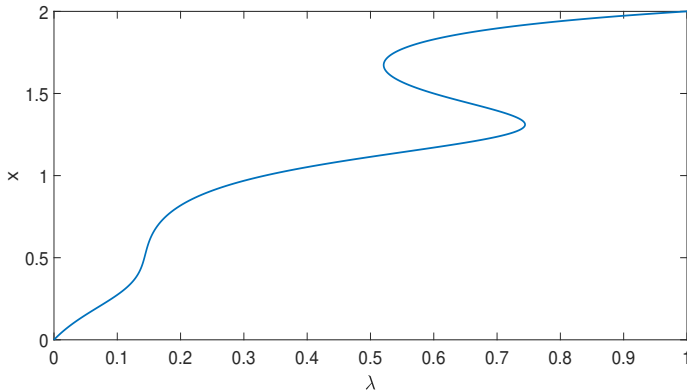
## Simple Continuation: Solution Set



**Note:** The plot shows the curve  $c := \{(x, \lambda) : H(x, \lambda) = 0\}$ ,  
i.e. **the solution set!**

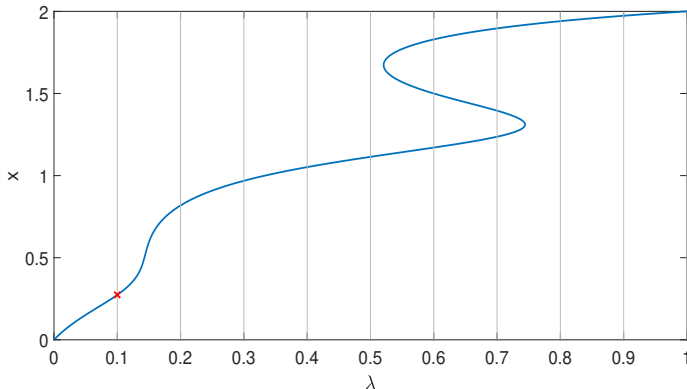
## Where Simple Continuation Fails

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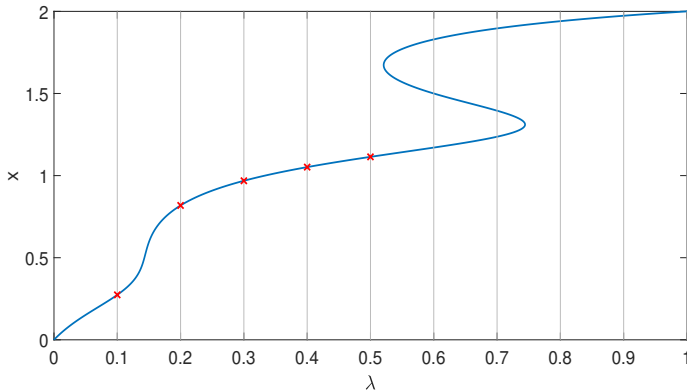
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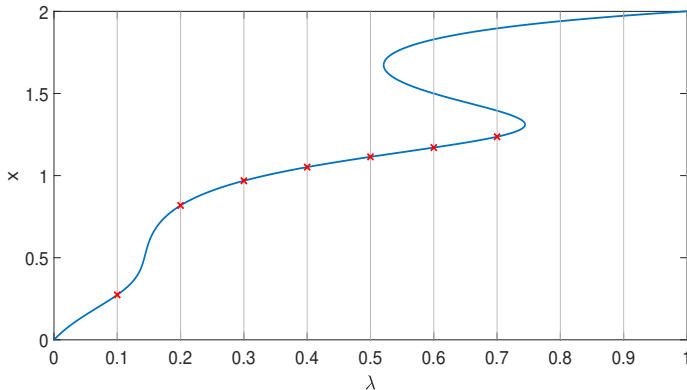
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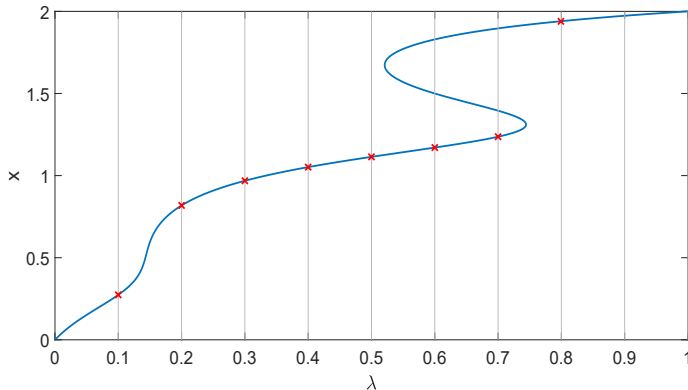
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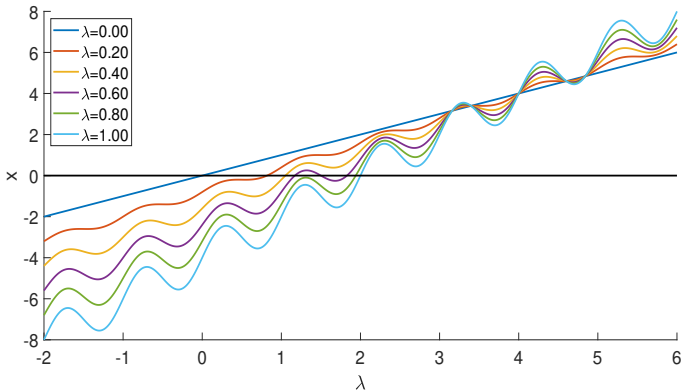
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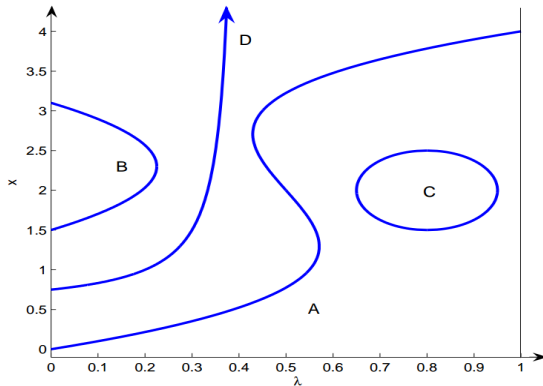


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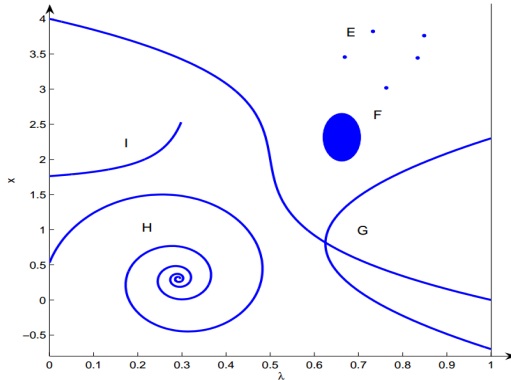
## Illustration of Possible Regular Solution Sets



Source: Borkovsky et al. [2010].



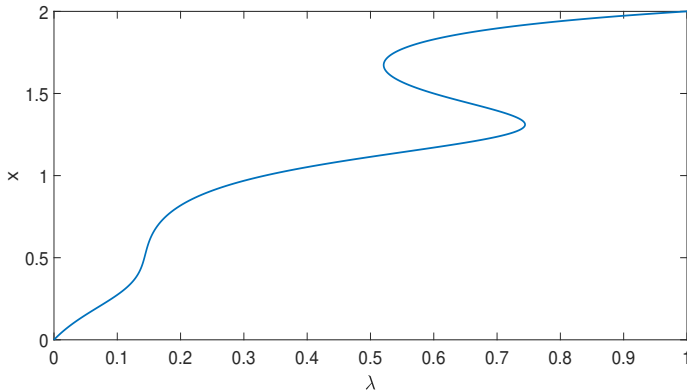
# Illustration of Possible Non Regular Solution Sets



Source: Borkovsky et al. [2010].

# Towards Predictor Corrector Methods

$$c := \{(x, \lambda) : H(x, \lambda) = 0\}$$



# Towards Predictor Corrector Methods

**Objective** Find the root of  $F(x) = 0$  by tracing the curve  
 $c := \{(x, \lambda) : H(x, \lambda) = 0\}.$

**Approach** Use the **arclength**  $s$  as parameterisation for the curve  $c$ .  
 $\Rightarrow$  The homotopy map changes to  $H(x(s), \lambda(s)) = 0!$

# Towards Predictor Corrector Methods: ODE-Theory

**Objective** Find the root of  $F(x) = 0$  by tracing the curve  
 $c := \{(x, \lambda) : H(x(s), \lambda(s)) = 0\}.$

- Differentiating  $H(x(s), \lambda(s))$  w.r.t.  $s$ , yields the initial and boundary value problem (IBVP)

$$x(0) = x_0, \quad \lambda(0) = 0, \quad \|(x'(s), \lambda'(s))\|_2^2 = 1, \quad (1)$$

$$\frac{\partial H(x(s), \lambda(s))}{\partial x} x'(s) + \frac{\partial H(x(s), \lambda(s))}{\partial \lambda} \lambda'(s) = 0. \quad (2)$$

- ODE-theory algorithms can solve the IBVP (1) - (2) to follow the curve  $c$  closely.

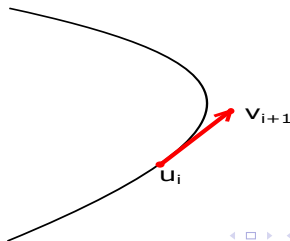
# Predictor Corrector Methods: Algorithm

**Approach** Trace  $c$  by *alternating* **prediction** and **correction** steps.

**Predictor** Use e.g., Euler's explicit step to predict

$$v_{i+1} = u_i + h \cdot H'(x(s_i), \lambda(s_i)).$$

**Corrector** Use the predicted point  $v_{i+1}$  and improve prediction by e.g., Newton-type methods.



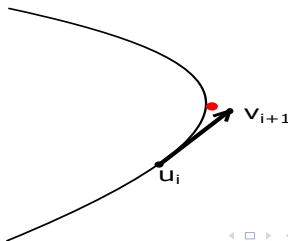
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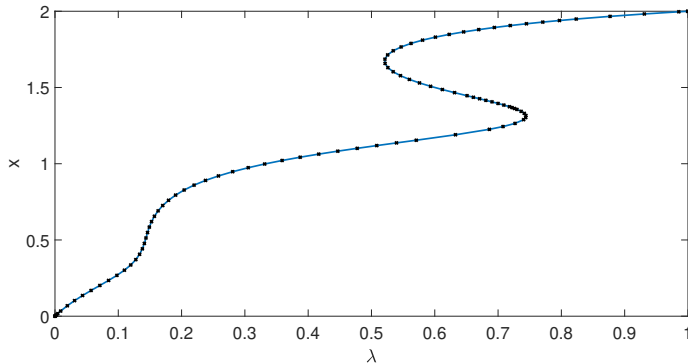
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# Predictor Corrector Methods: ODE-based Algorithm

$$H(x, \lambda) = (1 - \lambda)x + \lambda(2x - 4 + \sin(2\pi x))$$



Source: M-Hompack

# Probability One Globally Convergent Homotopy Methods

Given:  $H(x, \lambda, a) = (1 - \lambda)(x - a) + \lambda f(x)$

## Theorem

Let  $f : \mathbb{R}^p \rightarrow \mathbb{R}^p$  be a  $C^2$  map,  $H : \mathbb{R}^p \times [0, 1] \times \mathbb{R}^p \rightarrow \mathbb{R}^p$  a  $C^2$  map.

Suppose that

1)  $H$  is transversal to zero. (=full rank on  $H^{-1}(0)$ )

Suppose also that for each fixed  $a \in \mathbb{R}^p$ ,

2)  $H(0, x) = 0$  has a unique nonsingular solution  $x_0$ ,

3)  $H(1, x) = f(x)$  for  $\forall x \in \mathbb{R}^p$ .

4)  $H_a^{-1}(0)$  is bounded,

then  $H$  reaches a point  $(1; x)$  such that  $f(x) = 0$ . Furthermore, if  $Df(\bar{x})$  is invertible, then  $H^{-1}(0)$  has finite arc length.



# Competitive General Equilibrium

Goods  $j = 1, \dots, D$  (subscripts)

Prices  $(p_1, \dots, p_D)$

Agents  $i = 1, \dots, I$  (superscripts)

Endowment  $(w_1^i, \dots, w_D^i)$

Utility  $u^i$

Agent  $i$  solves her utility maximization problem

$$\max_{x^i} u(x^i) \tag{3}$$

$$\text{s.t. } px^i = pw^i \tag{4}$$

# Competitive General Equilibrium

For each of the  $I$  agents, we derive  $D$  FOCs

$$\frac{\partial u^i(x^i)}{\partial x_j^i} - \lambda^i p_j = 0 \quad i = 1, \dots, I, j = 1, \dots, D \quad (5)$$

The derivatives w.r.t. the lagrangian multipliers yield the  $I$  budget constraints from above

Market clearing must hold

$$\sum_{i=1}^I x_j^i - w_j^I = 0 \quad j = 1, \dots, D \quad (6)$$

Simplex normalization  $\sum_{j=1}^D p_j = 1$ .

System of nonlinear equations with  $ID + I + D$  equations and  $ID + I + D$  unknowns

- consumption allocations  $x_j^i$
- Lagrange multipliers  $\lambda^i$
- prices  $p_j$

# Parameter Continuation

**Objective** Solve parameterised non-linear equations of type  $F(x, \alpha) = 0$  for  $\alpha \in [a, b]$ .

$\alpha$  could be e.g. the *discount factor* in your model.

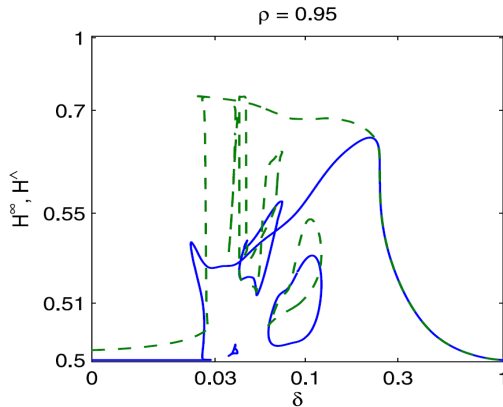
**Approach** Apply homotopy methods by defining the homotopy map as

$$H(x, \lambda) := F(x, \lambda) = 0,$$

from  $F(x, a)$  to  $F(x, b)$ .

**Benefit:** Predictor corrector methods find "all" solutions for  $\lambda \in [a, b]$ .

# Parameter Continuation



Parameter continuation in Doraszelski et al. [2010].

## Recap: Homotopy Algorithms in Economics

- We can use homotopy algorithms to **solve nonlinear systems of equations**  
by deforming a *globally solvable* system of equations  $g(x)$  into our target system of equations  $F(x)$
- Other applications include e.g.,
  - **solving parameterized families of nonlinear equations** depending on a single parameter, and
  - finding all solutions to polynomial equations by Judd et al. [2012],
  - in engineering: solving nonlinear programming problems, where optimizer fail [WATSON, 1999].

# HOMPACK90

- A collection of predictor corrector methods implemented in Fortran 90 by Watson et al. [1997]
- First application in economics by Schmedders [1998] for solving for equilibria in GE models with incomplete asset markets
- Well-established algorithms which take care of e.g.,
  - ① adaptive step sizes, and
  - ② an efficient implementation of the corrector step.
- Required user input: homotopy map  $H$  and its Jacobian  $JH$  as **Fortran subroutines**

# M-Hompack

- We have implemented and provide M-Hompack, an interface between Matlab and HOMPCK90
  - We chose to provide a Matlab interface for the beginning as it is widely used as modeling language
  - Researchers do not need to implement **any** Fortran code, but can implement the homotopy map and its Jacobian within Matlab
- ⇒ The application of homotopy methods becomes more feasible and even straight-forward



## Employed Computational Features

- Automatic differentiation
  - AD algorithms transform source codes of functions to their derivative
  - ⇒ **Homotopy methods rely on Jacobians; AD eliminates the need for any analytic differentiation**
- Sparsity
  - (If applicable) support of existing sparsity patterns in the Jacobian

# Counterfactual Analysis in Dynamic Models

**Objective** Evaluate the effect of a parameter change (= *policy change*) on the equilibrium.

- In models with **multiple equilibria**, the "correct" equilibrium is not well-defined without further information
  - In **factual scenarios** observe data select the equilibrium by maximum likelihood estimation.
  - For **counterfactual scenario**, we cannot observe these data.
- There exists **no** well-established algorithm for counterfactual analysis.

# Toy Model Aguirregabiria [2012], Modified.

The fixed-point equation

$$\psi(P, \beta) = h(\beta)G(P) \quad (7)$$

defines the equilibrium  $P$  implicitly.  $G$  denotes the Cauchy cumulative density function (CDF)

$$G(P) = \frac{1}{\pi} \operatorname{atan} \left( \frac{P - 0.5}{0.15} \right) + 0.5, \quad (8)$$

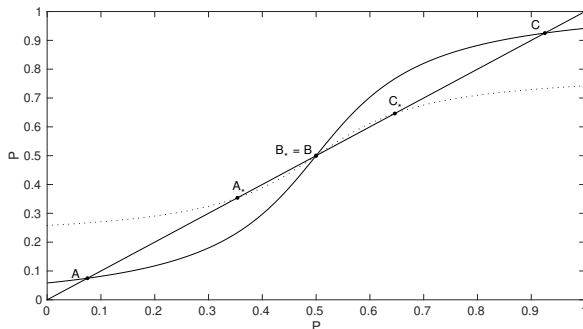
and

$$h(\beta) = \frac{1}{3\beta^5 + 3}. \quad (9)$$

$\hat{\beta}_0 := 0.85$  denotes the value of  $\beta$  in the factual scenario, and  $\beta_* := 1.15$  in the **counterfactual** scenario

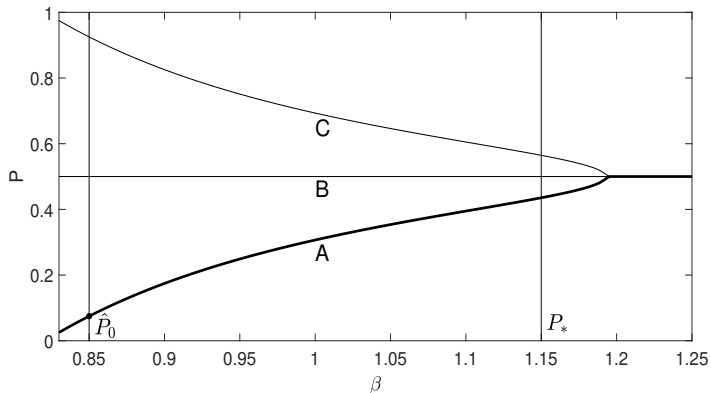
# Fixed-Point Plot

$$\psi(P, \beta) = h(\beta)G(P)$$



$\beta = \hat{\beta}_0$  (solid line) and  $\beta = \beta_*$  (dotted line) with the equilibria  $\{A, B, C\}$

# Equilibria Paths: As Function of $\beta$



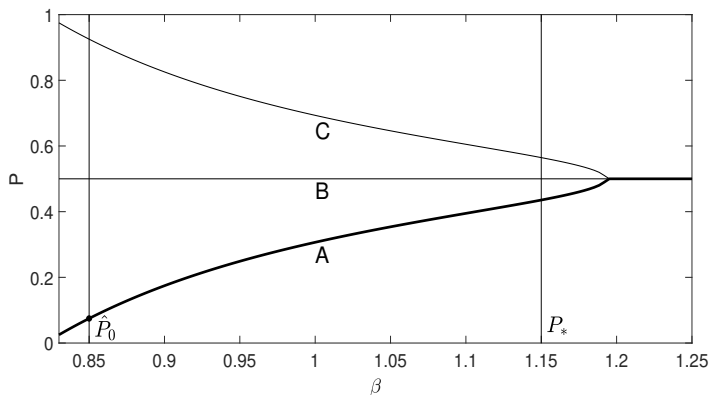
# Counterfactual Analysis following Aguirregabiria [2012]

**Assume** Invariance of equilibrium type (IET) by Aguirregabiria [2012]

*Definition (IET):* The factual equilibrium  $\hat{P}_0$  and the counterfactual equilibrium  $P_*$  are connected by a continuous path.

**Idea** The fixed-point equation for  $\hat{P}_0$  and for  $P_*$  are connected by a continuous path. **Trace the path** by varying the **parameter of interest**  $\beta$ .

# Equilibria Paths: As a Function of $\beta$



**Note:** We can trace these paths with our homotopy methods!

# The Homotopy Algorithm

**Objective** Find the counterfactual equilibrium "*of same type*" for your chosen counterfactual parameter  $\beta_*$ .

**Step 1** Build the homotopy map

$$H(P, \lambda) := \psi(P, (1 - \lambda)\beta_0 + \lambda\beta_*) - P \quad (10)$$

from the factual scenario  $\hat{P}_0$  to the counterfactual scenario  $P_*$ .

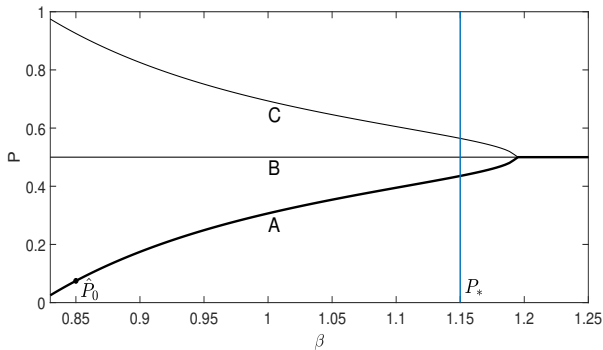
**Step 2** Follow the curve  $c$  by employing the ODE predictor corrector method.

**Step 3** For  $\lambda = 1$  and  $H(\bar{P}, 1) = 0$ :  $\bar{P}$  equals our counterfactual equilibrium  $P_*$ .



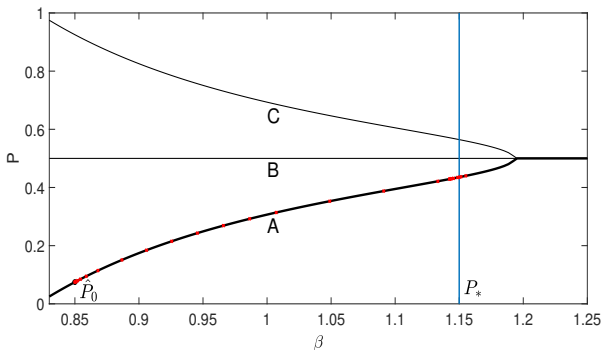
# Result M-Hompack

$$\hat{\beta}_0 = 0.85 \text{ and } \beta_* = 1.15$$



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**Note:** M-Hompack solves for the "correct" counterfactual equilibrium, and even yields intermediate equilibria as byproduct!

# Aguirregabiria [2012]'s Algorithm

## Simple Homotopy Algorithm

**Objective** Find the counterfactual equilibrium "*of same type*" for your chosen counterfactual parameter  $\beta_*$ .

**Step 1** Derive the **first-order Taylor expansion**  $F_T$  around the factual equilibrium  $F(\beta_0) := \hat{P}_0$  as

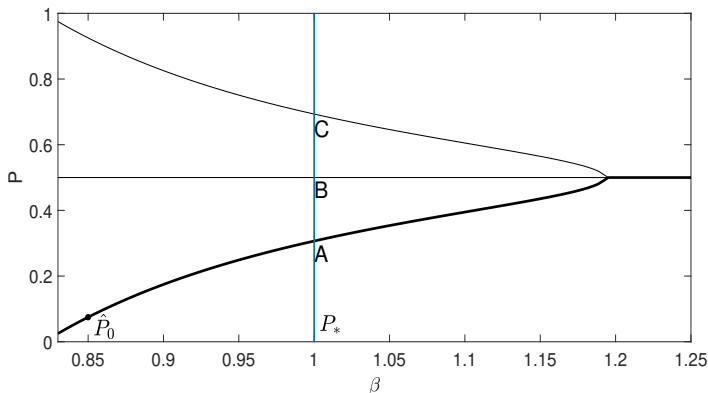
$$F_T(\beta) = F(\hat{\beta}_0) + \frac{\partial F(\hat{\beta}_0)}{\partial \beta}(\beta - \hat{\beta}_0)$$

**Step 2** Approximate the counterfactual equilibrium  $P_*$  by  $\tilde{P}_* = F_T(\beta_*)$

**Step 3** Use  $\tilde{P}_*$  as starting point for the equilibrium mapping  $\psi(P, \beta)$  and iterate up to convergence.

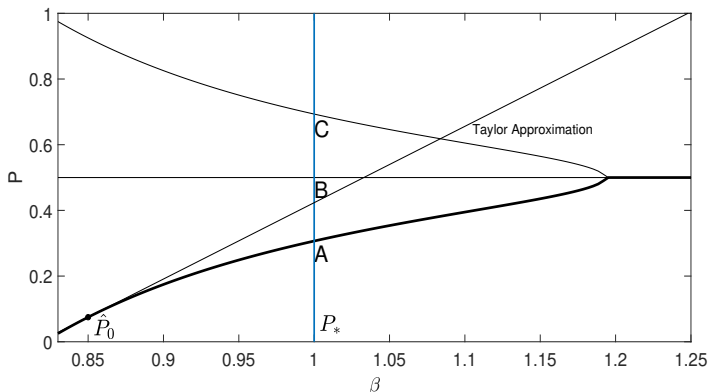
# Simple Homotopy Algorithm Aguirregabiria [2012]

$$\hat{\beta}_0 = 0.85 \text{ and } \beta_* = 1.0$$



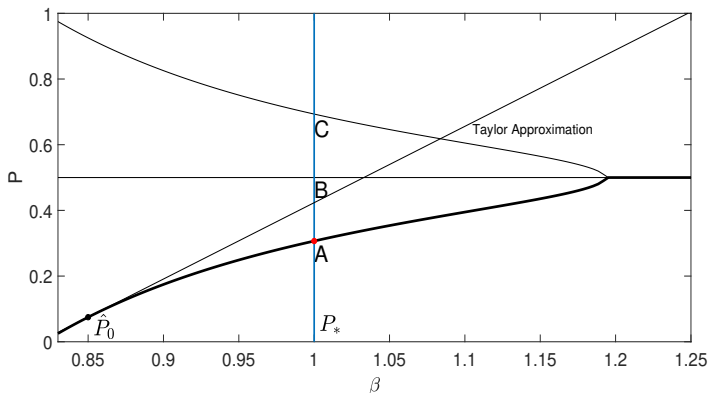
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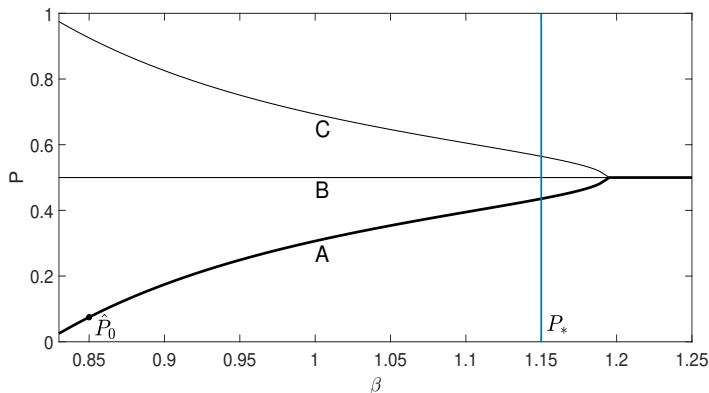
# Simple Homotopy Algorithm Aguirregabiria [2012]

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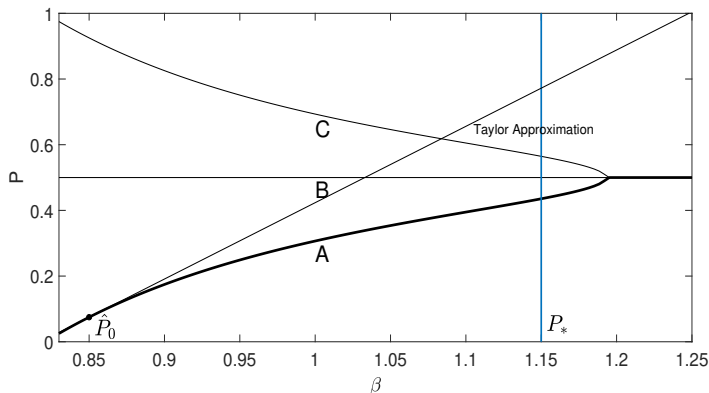
# Simple Homotopy Algorithm Aguirregabiria [2012]

$$\hat{\beta}_0 = 0.85 \text{ and } \beta_* = 1.15$$



# Simple Homotopy Algorithm Aguirregabiria [2012]

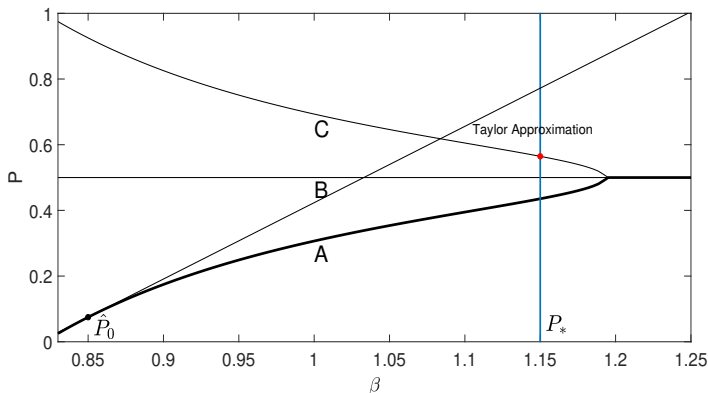
$$\hat{\beta}_0 = 0.85 \text{ and } \beta_* = 1.15$$





# Simple Homotopy Algorithm Aguirregabiria [2012]

$$\hat{\beta}_0 = 0.85 \text{ and } \beta_* = 1.15$$



# Live-Demonstration

- Matlab Live-Demo

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