

# Viscosity Solutions for Dummies (including Economists)

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Online Appendix to

“Income and Wealth Distribution in Macroeconomics:  
A Continuous-Time Approach”

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# Viscosity Solutions

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For our purposes, useful for two reasons:

1. problems with kinks, e.g. coming from non-convexities
2. problems with “state constraints”
  - borrowing constraints
  - computations with bounded domain

Things to remember

1. viscosity solution  $\Rightarrow$  **no concave kinks** (convex kinks are allowed)
2. **uniqueness**: HJB equations have **unique** viscosity solution
3. **constrained** viscosity solution: “**boundary inequalities**”

# References

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Relatively accessible ones:

- Lions (1983) “Hamilton-Jacobi-Bellman Equations and the Optimal Control of Stochastic Systems” – report of some results, no proofs  
<http://www.mathunion.org/ICM/ICM1983.2/Main/icm1983.2.1403.1418.ocr.pdf>
- Crandall (1995) “Viscosity Solutions: A Primer”  
<http://www.princeton.edu/~moll/crandall-primer.pdf>
- Bressan (2011) “Viscosity Solutions of Hamilton-Jacobi Equations and Optimal Control Problems”  
<https://www.math.psu.edu/bressan/PSPDF/hj.pdf>
- Yu (2011) <http://www.math.ualberta.ca/~xinweiyu/527.1.08f/lec11.pdf>

Less accessible but often cited:

- Crandall, Ishii and Lions (1992) “User’s Guide to Viscosity Solutions”
- Bardi and Capuzzo-Dolcetta (2008)  
<https://www.dropbox.com/s/hsihfm8xwnncvy/bardi-capuzzo-wholebook.pdf?dl=0>
- Fleming & Soner (2006)  
<https://www.dropbox.com/s/wbekmg2icp5u9i1/fleming-soner-wholebook.pdf?dl=0>

# Outline

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1. Definition of viscosity solution
2. No concave kinks
3. Constrained viscosity solutions
4. Uniqueness

# Viscosity Solutions: Definition

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Consider HJB for generic optimal control problem

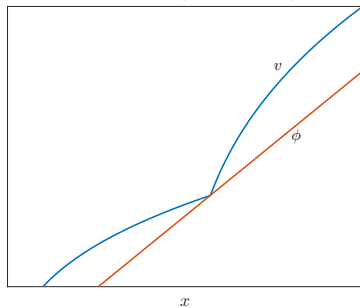
$$\rho v(x) = \max_{\alpha \in A} \{r(x, \alpha) + v'(x)f(x, \alpha)\} \quad (\text{HJB})$$

- Next slide: definition of viscosity solution
- Basic idea:  $v$  may have kinks i.e. may not be differentiable
- replace  $v'(x)$  at point where it does not exist (because of kink in  $v$ ) with derivative of smooth function  $\phi$  touching  $v$
- two types of kinks: concave and convex  $\Rightarrow$  two conditions
  - **concave** kink:  $\phi$  touches  $v$  **from above**
  - **convex** kink:  $\phi$  touches  $v$  **from below**
- Remark: definition allows for concave kinks. In a few slides: these never arise in **maximization problems**

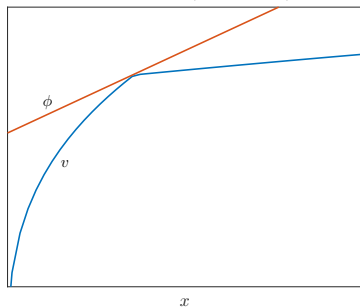
# Convex and Concave Kinks

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Convex Kink (Supersolution)



Concave Kink (Subsolution)



# Viscosity Solutions: Definition

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**Definition:** A **viscosity solution** of (HJB) is a continuous function  $v$  such that the following hold:

1. (Subsolution) If  $\phi$  is any smooth function and if  $v - \phi$  has a local maximum at point  $x^*$  ( $v$  may have a concave kink), then

$$\rho v(x^*) \leq \max_{\alpha \in A} \{r(x^*, \alpha) + \phi'(x^*)f(x^*, \alpha)\}$$

2. (Supersolution) If  $\phi$  is any smooth function and if  $v - \phi$  has a local minimum at point  $x^*$  ( $v$  may have a convex kink), then

$$\rho v(x^*) \geq \max_{\alpha \in A} \{r(x^*, \alpha) + \phi'(x^*)f(x^*, \alpha)\}.$$

## A few remarks, terminology

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- If  $v$  is differentiable at  $x^*$ , then
  - local max or min of  $v - \phi$  implies  $v'(x^*) = \phi'(x^*)$
  - sub- and supersolution conditions  $\Rightarrow$  viscosity solution of (HJB) is just classical solution
- If a continuous function  $v$  satisfies condition 1 (but not necessarily 2) we say that it is a “viscosity subsolution”
- Conversely, if  $v$  satisfies condition 2 (but not necessarily 1), we say that it is a “viscosity supersolution”
- $\Leftrightarrow$  a continuous function  $v$  is a “viscosity solution” if it is both a “viscosity subsolution” and a “viscosity supersolution.”
- “subsolution” and “supersolution” come from  $\leq 0$  and  $\geq 0$
- “viscosity” is in honor of the “method of vanishing viscosity”: add Brownian noise and  $\rightarrow 0$  (movements in viscous fluid)



# Viscosity Solutions: Intuition

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- Consider discrete time Bellman:

$$v(x) = \max_{\alpha} r(x, \alpha) + \beta v(x'), \quad x' = f(x, \alpha)$$

- Think about it as an operator  $T$  on function  $v$

$$(Tv)(x) = \max_{\alpha} r(x, \alpha) + \beta v(f(x, \alpha))$$

- Solution = fixed point:  $Tv = v$
- Intuitive property of  $T$  (“monotonicity”)

$$\phi(x) \leq v(x) \quad \forall x \quad \Rightarrow \quad (T\phi)(x) \leq (Tv)(x) \quad \forall x \quad (*)$$

- Intuition: if my continuation value is higher, I’m better off
- **Viscosity solution is exactly same idea**
- Key idea: sidestep non-differentiability of  $v$  by using “monotonicity”

# Viscosity Solutions: Heuristic Derivation

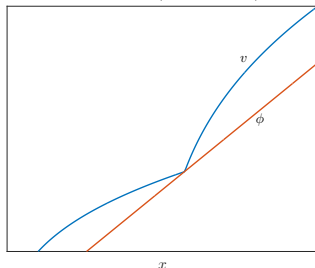
- Time periods of length  $\Delta$ . Consider HJB equation

$$v(x_t) = \max_{\alpha} \Delta r(x_t, \alpha) + (1 - \rho\Delta)v(x_{t+\Delta}) \quad \text{s.t.}$$

$$x_{t+\Delta} = \Delta f(x_t, \alpha) + x_t$$

- Suppose  $v$  is not differentiable at  $x^*$  and has a **convex kink**
  - problem: when taking  $\Delta \rightarrow 0$ , pick up derivative  $v'$
  - solution: replace continuation value with smooth function  $\phi$

Convex Kink (Supersolution)



# Viscosity Solutions: Heuristic Derivation

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- For now: consider  $\phi$  such that  $\phi(x^*) = v(x^*)$ 
  - local min of  $v - \phi$  and  $\phi(x^*) = v(x^*) \Rightarrow v(x) > \phi(x), x \neq x^*$
- Then for  $x_t = x^*$ , we have

$$v(x_t) \geq \max_{\alpha} \Delta r(x_t, \alpha) + (1 - \rho\Delta)\phi(x_{t+\Delta})$$

- Subtract  $(1 - \rho\Delta)\phi(x_t)$  from both sides and use  $\phi(x_t) = v(x_t)$

$$\Delta\rho v(x_t) \geq \max_{\alpha} \Delta r(x_t, \alpha) + (1 - \rho\Delta)(\phi(x_{t+\Delta}) - \phi(x_t))$$

- Dividing by  $\Delta$  and letting  $\Delta \rightarrow 0$  yields the **supersolution condition**

$$\rho v(x_t) \geq \max_{\alpha} r(x_t, \alpha) + \phi'(x_t)f(x_t, \alpha)$$

# Viscosity Solutions: Heuristic Derivation

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- Turns out this works for **any**  $\phi$  such that  $v - \phi$  has local min at  $x^*$ 
  - define  $\kappa = v(x^*) - \phi(x^*)$ . Then  $v(x^*) = \phi(x^*) + \kappa$  and

$$v(x_t) \geq \max_{\alpha} \Delta r(x_t, \alpha_t) + (1 - \rho\Delta)(\phi(x_{t+\Delta}) + \kappa)$$

- subtract  $(1 - \rho\Delta)(\phi(x_t) + \kappa)$  from both sides

$$\Delta\rho v(x_t) \geq \max_{\alpha} \Delta r(x_t, \alpha) + (1 - \rho\Delta)(\phi(x_{t+\Delta}) - \phi(x_t))$$

- rest is the same...
- Derivation of **subsolution condition** exactly **symmetric**

# Viscosity Solution: Nonstandard but Intuitive Definition

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Monotonicity logic: if my continuation value is higher, I'm better off

Suggests alternative definition that is less standard but more intuitive and in particular let's you remember which way  $\leq$  and  $\geq$  go

**Definition:** A **viscosity solution** of (HJB) is a continuous function  $v$  such that the following hold:

1. (Subsolution) If  $\phi$  is any smooth function that touches  $v$  **from above** at point  $x^*$  ( $v$  may have a concave kink & " **$\phi$  is better than  $v$** "), then

$$\rho v(x^*) \leq \max_{\alpha \in A} \{r(x^*, \alpha) + \phi'(x^*)f(x^*, \alpha)\}$$

Intuition: analogue of  $\phi \geq v \Rightarrow T\phi \geq Tv$  for discrete-time Bellman

2. (Supersolution) If  $\phi$  is any smooth function that touches  $v$  **from below** at point  $x^*$  ( $v$  may have a convex kink & " **$\phi$  is worse than  $v$** "), then

$$\rho v(x^*) \geq \max_{\alpha \in A} \{r(x^*, \alpha) + \phi'(x^*)f(x^*, \alpha)\}$$

Intuition: analogue of  $\phi \leq v \Rightarrow T\phi \leq Tv$  for discrete-time Bellman

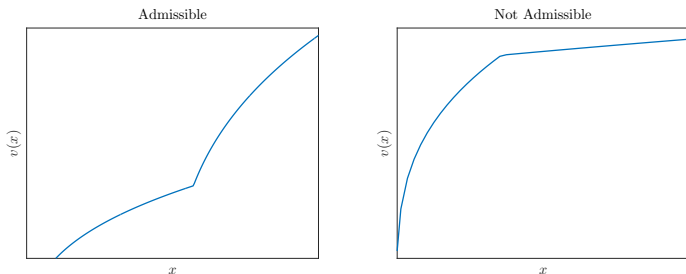
# Viscosity + maximization $\Rightarrow$ no concave kinks

## Proposition

*The viscosity solution of the maximization problem*

$$\rho v(x) - \max_{\alpha \in A} \{ r(x, \alpha) + v'(x) f(x, \alpha) \} = 0$$

*only admits convex (downward) kinks, but not concave (upward) kinks.*



- opposite would be true for minimization problem

# Why is this useful? Problems with non-covexities

- Consider growth model

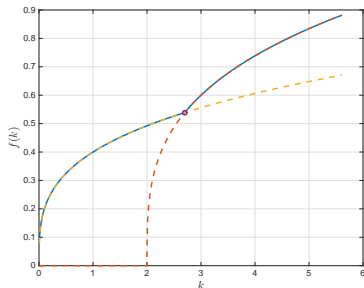
$$\rho v(k) = \max_c u(c) + v'(k)(F(k) - \delta k - c).$$

- But drop assumption that  $F$  is strictly concave. Instead: “butterfly”

$$F(k) = \max\{F_L(k), F_H(k)\},$$

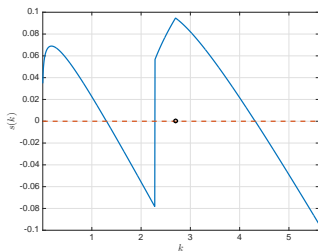
$$F_L(k) = A_L k^\alpha,$$

$$F_H(k) = A_H((k - \kappa)^+)^{\alpha}, \quad \kappa > 0, A_H > A_L$$

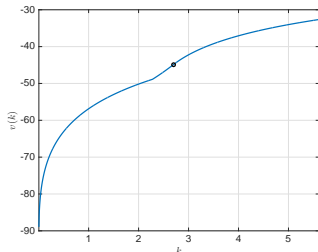


# Why is this useful? Problems with non-covexities

- Suppose restrict attention to solutions  $v$  with **at most one kink**
  - there is solution we found in Lecture 3 with **convex kink**



(a) Saving Policy Function



(b) Value Function

- but there is **another solution** with a **concave kink**!
- Proposition above **rules out** second solution



# Proof that no concave kinks

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- **Step 1:** maximization problem  $\Rightarrow$  convex Hamiltonian
- **Step 2:** convex Hamiltonian  $\Rightarrow$  no concave kinks
- Proof of Step 1: Hamiltonian is

$$H(x, p) := \max_{\alpha \in A} \{r(x, \alpha) + pf(x, \alpha)\}$$

- first- and second-order conditions:

$$r_{\alpha} + pf_{\alpha} = 0 \quad (\text{FOC})$$

$$r_{\alpha\alpha} + pf_{\alpha\alpha} \leq 0 \quad (\text{SOC})$$

- From (FOC)

$$\alpha_p = -\frac{f_{\alpha}}{(r_{\alpha\alpha} + pf_{\alpha\alpha})}$$

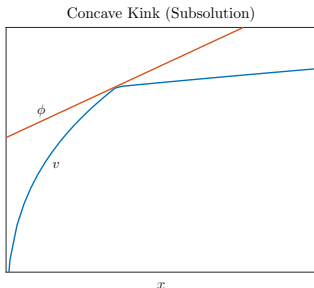
- Differentiating Hamiltonian, we have  $H_p(x, p) = f(x, \alpha(x, p))$  and

$$H_{pp} = f_{\alpha}\alpha_p = -\frac{(f_{\alpha})^2}{(r_{\alpha\alpha} + pf_{\alpha\alpha})} > 0$$

## Proof that no concave kinks (step 2)

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- Proof by contradiction: suppose concave kink at  $x^*$
- Then  $v, \phi$  in subsolution condition look like in figure



- Note: left & right derivatives satisfy

$$\partial_- v(x^*) < \phi'(x^*) < \partial_+ v(x^*)$$

- Hence there exists  $t \in (0, 1)$  such that

$$\phi'(x^*) = t\partial_+ v(x^*) + (1 - t)\partial_- v(x^*)$$

## Proof that no concave kinks (step 2)

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- Because  $H$  is convex, for  $t$  defined on previous slide

$$H(x^*, \phi'(x^*)) < tH(x^*, \partial_+ v'(x^*)) + (1 - t)H(x^*, \partial_- v'(x^*)) \quad (*)$$

- By continuity of  $v$

$$\rho v(x^*) = H(x^*, \partial_+ v(x^*)),$$

$$\rho v(x^*) = H(x^*, \partial_- v(x^*))$$

- Therefore (\*) implies

$$H(x^*, \phi'(x^*)) < \rho v(x^*)$$

- But this contradicts the subsolution condition

$$\rho v(x^*) \leq H(x^*, \phi'(x^*)). \quad \square$$

# Something to Keep in Mind

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- Viscosity solution **can handle** problems where  $v$  has **kinks**...
- ... but **not discontinuities**
- Theory can be extended in special cases, but no general theory of discontinuous viscosity solutions
- In economics, kinks seem more common than discontinuities
- Good reference on HJB equations with discontinuities: Barles and Chasseigne (2015) “(Almost) Everything You Always Wanted to Know About Deterministic Control Problems in Stratified Domains”  
<http://arxiv.org/abs/1412.7556>

# Viscosity Solutions with State Constraints

# Constrained Viscosity Soln: Boundary Inequalities

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- How handle “state constraints”?
  - borrowing constraints
  - computations with bounded domain
- Example: growth model with state constraint

$$v(k_0) = \max_{\{c(t)\}_{t \geq 0}} \int_0^{\infty} e^{-\rho t} u(c(t)) dt \quad \text{s.t.}$$

$$\dot{k}(t) = F(k(t)) - \delta k(t) - c(t)$$

$$k(t) \geq k_{\min} \text{ all } t \geq 0$$

- purely pedagogical: constraint will never bind if  $k_{\min} < \text{st.st.}$
- HJB equation

$$\rho v(k) = \max_c \{u(c) + v'(k)(F(k) - \delta k - c)\} \quad (\text{HJB})$$

- Key question: **how impose  $k \geq k_{\min}$ ?**

## Example: Growth Model with Constraint

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- **Result:** if  $v$  is (left-)differentiable at  $k_{\min}$ , it needs to satisfy

$$v'(k_{\min}) \geq u'(F(k_{\min}) - \delta k_{\min}) \quad (\text{BI})$$

- **Intuition:**

- $v'(k_{\min})$  is such that if  $k(t) = k_{\min}$  then  $\dot{k}(t) \geq 0$
- if  $v$  is differentiable, the FOC still holds at the constraint

$$u'(c(k_{\min})) = v'(k_{\min}) \quad (\text{FOC})$$

- for constraint not to be violated, need

$$F(k_{\min}) - \delta k_{\min} - c(k_{\min}) \geq 0 \quad (*)$$

- (FOC) and (\*)  $\Rightarrow$  (BI).
- Next: state constraints in generic optimal control problem

# Generic Control Problem with State Constraint

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- Consider variant of generic maximization problem

$$v(x) = \max_{\{\alpha(t)\}_{t \geq 0}} \int_0^{\infty} e^{-\rho t} r(x(t), \alpha(t)) dt \quad \text{s.t.}$$
$$\dot{x}(t) = f(x(t), \alpha(t)), \quad x(0) = x$$
$$x(t) \geq x_{\min} \quad \text{all } t \geq 0$$

- HJB equation

$$\rho v(x) = \max_{\alpha \in A} \{r(x, \alpha) + v'(x)f(x, \alpha)\} \quad (\text{HJB})$$

- Question: **how impose  $x \geq x_{\min}$ ?** Two cases:
  - $v$  is (left-)differentiable at  $x_{\min}$ : boundary inequality
  - $v$  **not** differentiable at  $x_{\min}$ : “constrained viscosity solution”



# State constraints if $v$ is differentiable

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- Use Hamiltonian formulation

$$\rho v(x) = H(x, v'(x)) \quad (\text{HJB})$$

$$H(x, p) := \max_{\alpha \in A} \{r(x, \alpha) + pf(x, \alpha)\} \quad (\text{H})$$

- From envelope condition

$$H_p(x, v'(x)) = f(x, \alpha^*(x)) = \text{optimal drift at } x$$

- If  $v'$  and  $H_p$  exist, (HJB) for generic control problem satisfies

$$H_p(x_{\min}, v'(x_{\min})) \geq 0 \quad (\text{BI})$$

- see Soner (1986, p.553) and Fleming and Soner (2006, p.108)
  - write state constraint as  $f(x, \alpha^*(x)) \cdot \nu(x) \leq 0$  for  $x$  at boundary where  $\nu(x)$  = “outward normal vector” at boundary
  - show this implies  $H_p(x, \nabla v(x)) \cdot \nu(x) \geq 0$

## If $v$ not differentiable: constrained viscosity solution

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- Setting  $H_p(x_{\min}, v'(x_{\min})) \geq 0$  obviously requires  $v$  to be (left-)differentiable  $\Rightarrow$  what if not differentiable at  $x_{\min}$ ?
- **Definition:** a **constrained viscosity solution** of (HJB) is a continuous function  $v$  such that
  1.  $v$  is a viscosity solution (i.e. both sub- and supersolution) for all  $x > x_{\min}$
  2.  $v$  is a **subsolution** at  $x = x_{\min}$ : if  $\phi$  is any smooth function and if  $v - \phi$  has a local maximum at point  $x_{\min}$ , then

$$\rho v(x_{\min}) \leq \max_{\alpha \in A} \{r(x_{\min}, \alpha) + \phi'(x_{\min})f(x_{\min}, \alpha)\} \quad (*)$$

- (\*) functions as **boundary condition**, or rather “**boundary inequality**”
- Note: minimization  $\Rightarrow$  opposite, i.e. **supersolution** at boundary
  - for minimization: Fleming and Soner (2006), Section II.12
  - for maximization: e.g. Definition 3.1 in Zariphopoulou (1994)

## Intuition why subsolution on boundary

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- Follow Fleming and Soner (2006, p.108) with signs switched
- **Lemma:** If  $v$  is diff'ble at  $x_{\min}$ , then  $f(x_{\min}, \alpha^*(x_{\min})) \geq 0$  implies

$$H(x_{\min}, p) \geq H(x_{\min}, v'(x_{\min})), \quad \text{for all } p \geq v'(x_{\min}) \quad (\text{Bl}')$$

- Proof: for any  $p \leq v'(x_{\min})$

$$\begin{aligned} H(x_{\min}, p) &= \max_{\alpha \in A} \{h(x_{\min}, \alpha) + pf(x_{\min}, \alpha)\} \\ &\geq r(x_{\min}, \alpha^*(x_{\min})) + pf(x_{\min}, \alpha^*(x_{\min})) \\ &\geq r(x_{\min}, \alpha^*(x_{\min})) + v'(x_{\min})f(x_{\min}, \alpha^*(x_{\min})) \\ &= H(x_{\min}, v'(x_{\min})). \quad \square \end{aligned}$$

- Remark: (Bl') implies (Bl), i.e.  $H_p(x_{\min}, v'(x_{\min})) \geq 0$

## Intuition why subsolution on boundary

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- By continuity of  $v$  and  $v'$ :

$$\rho v(x_{\min}) = H(x_{\min}, v'(x_{\min}))$$

- Combining with (BI'):

$$\rho v(x_{\min}) \leq H(x_{\min}, p) \quad \text{for all } p \geq v'(x_{\min})$$

- **Subsolution condition** same statement without differentiability
  - if  $v$  (left-)differentiable: local max of  $v - \phi$  at  $x_{\min}$   
 $\Leftrightarrow v'(x_{\min}) \leq \phi'(x_{\min})$  (not = bc/ corner)
  - but also applies if  $v$  not differentiable

## Example: Growth Model with Bounded Domain

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- Consider again HJB equation for growth model

$$\rho v(k) = \max_c \{u(c) + v'(k)(F(k) - \delta k - c)\}$$

- For numerical solution, want to impose

$$k_{\min} \leq k(t) \leq k_{\max} \quad \text{all } t$$

- How can we ensure this?
- Answer: impose **two boundary inequalities**

$$v'(k_{\min}) \geq u'(F(k_{\min}) - \delta k_{\min})$$

$$v'(k_{\max}) \leq u'(F(k_{\max}) - \delta k_{\max})$$

# Uniqueness of Viscosity Solution

# Uniqueness of Viscosity Solution

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- **Theorem:** Under some conditions, HJB equation has unique viscosity solution
- due to Crandall and Lions (1983), “Viscosity Solutions of Hamilton-Jacobi Equations”
- My intuition for uniqueness theorem with state constraints in one dimension
  - ODE has unique solution given one boundary condition
  - two boundary inequalities = one boundary condition ...
  - ... sufficient to pin down unique solution
- but much more powerful: generalizes to  $N$  dimensions, kinks etc

# Intuition for Uniqueness: Boundary Inequalities

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- Consider toy problem with explicit solution

$$v(x) = \max_{\{\alpha(t)\}_{t \geq 0}} \int_0^{\infty} e^{-t} \left( -3x(t)^2 - \frac{1}{2}\alpha(t)^2 \right) dt,$$

$$\dot{x}(t) = \alpha(t) \quad x(0) = x,$$

and with state constraints  $x(t) \in [x_{\min}, x_{\max}]$

- HJB equation

$$v(x) = \max_{\alpha} \left\{ -3x^2 - \frac{1}{2}\alpha^2 + v'(x)\alpha \right\}$$

- or maximizing out  $\alpha$  using  $\alpha = v'(x)$

$$v(x) = -3x^2 + \frac{1}{2}(v'(x))^2 \quad (\text{HJB})$$

- “Correct” solution is (verify:  $-x^2 = -3x^2 + \frac{1}{2}(-2x)^2$ )

$$v(x) = -x^2$$

- Respects state constraints, e.g.  $\dot{x} = v'(x) = -2x < 0$  if  $x > 0$



## Intuition for Uniqueness: Boundary Inequalities

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- $\dot{x}(t) = v'(x(t)) \Rightarrow$  natural state constraint boundary conditions

$$v'(x_{\min}) \geq 0, \quad v'(x_{\max}) \leq 0 \quad (\text{BI})$$

- ensure that  $\dot{x} \geq 0$  at  $x = x_{\min}$  and  $\dot{x} \leq 0$  at  $x = x_{\max}$
- **Result:**  $v(x) = -x^2$  is **the only** continuously differentiable solution of (HJB) which satisfies the boundary inequalities (BI)
  - proof on next slide
- Result is striking because
  - inequalities (BI) strong enough to pin down unique solution of (HJB) even though, at correct solution  $v(x) = -x^2$ , **neither holds with equality**
  - boundary inequalities pin down unique solution even though this solution “does not see the boundaries”

# Proof of Result

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- Let  $v$  be smooth solution of (HJB). Let  $v(0) = c$ . Rule out  $c \neq 0$
- Easy to rule out  $c < 0$ : (HJB) at  $x = 0$  is  $c = \frac{1}{2}(v'(0))^2 \Rightarrow$  no solution for  $v'(0)$  when  $c < 0$
- Next consider  $c > 0$ . Inverting (HJB) for  $v'$  yields two branches

$$\text{Branch 1: } v'(x) = +\sqrt{6x^2 + 2v(x)}$$

$$\text{Branch 2: } v'(x) = -\sqrt{6x^2 + 2v(x)}$$

- Importantly, continuous  $v' \Rightarrow v$  satisfies either branch 1 or branch 2 for all  $x \in (x_{\min}, x_{\max})$ , i.e. cannot switch branches
  - in particular  $c > 0 \Rightarrow$  switching branches not allowed at  $x = 0$
- Suppose  $v$  satisfies branch 1 for all  $x$ . Then
  - $v'(0) = \sqrt{2c} > 0$  and  $v(x) > c$  for  $x > 0$  near  $x = 0$
  - Similarly  $v(x) > c, v'(x) > 0$  all  $x > 0 \Rightarrow$  violate  $v'(x_{\max}) \leq 0$
- Suppose  $v$  satisfies branch 2 for all  $x$ . Then  $v'(0) = -\sqrt{2c} < 0 \dots$ 
  - ...  $v(x) > c, v'(x) < 0$  all  $x < 0 \Rightarrow$  violate  $v'(x_{\min}) \geq 0 \square$

## Remark: strengthening the result

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- Have restricted attention to solutions such that  $v'$  is continuous
- In fact, result can be strengthened further to say
- **Result 2:**  $v(x) = -x^2$  is **the only** solution of (HJB) without concave kinks (“viscosity solution”) which satisfies boundary inequalities (BI)

# Uniqueness Theorem: Logic

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- Here outline case with state constraints:  $x \in [x_{\min}, x_{\max}]$ 
  - due to Soner (1986), Capuzzo-Dolcetta & Lions (1990)
  - use notation  $X := (x_{\min}, x_{\max})$  and  $\bar{X} := [x_{\min}, x_{\max}]$
  - can extend to unbounded domain  $x \in \mathbb{R}$  (references later)
- Key step in uniqueness proof: “comparison theorem”
- Consider HJB equation

$$\rho v(x) = \max_{\alpha \in A} \{r(x, \alpha) + v'(x)f(x, \alpha)\} \quad (\text{HJB})$$

with state constraints  $x \in \bar{X}$

- **Comparison theorem:** (Under certain assumptions...) if  $v_1$  is subsolution of (HJB) on  $\bar{X}$  &  $v_2$  is supersolution of (HJB) on  $X$ , then

$$v_1(x) \leq v_2(x) \quad \text{for all } x \in \bar{X}.$$

## Remark: another comparison theorem you may know

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- Here's another comparison theorem you may know that has same structure: **Grönwall's inequality**

- If  $v_1'(t) \leq \beta(t)v_1(t)$  then  $v_1(t) \leq v_1(0) \exp\left(\int_0^t \beta(s)ds\right)$

- This is really: If  $v_1'(t) \leq \beta(t)v_1(t)$  and  $v_2'(t) = \beta(t)v_2(t)$ , then

$$v_1(t) \leq v_2(t) \quad \text{for all } t$$

# Comparison Theorem immediately $\Rightarrow$ Uniqueness

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- **Corollary (uniqueness):** there exists a unique **constrained viscosity solution** of (HJB) on  $\bar{X}$ , i.e. if  $v_1$  and  $v_2$  are both constrained viscosity solutions of (HJB) on  $\bar{X}$ , then  $v_1(x) = v_2(x)$  for all  $x \in \bar{X}$
- **Proof:** let  $v_1, v_2$  be constrained viscosity solutions of (HJB) on  $\bar{X}$ 
  1. Since  $v_1$  and  $v_2$  are constrained viscosity solutions,  $v_1$  is also a **subsolution** on  $\bar{X}$  and  $v_2$  is a **supersolution** on  $X$ . By the comparison theorem, therefore  $v_1(x) \leq v_2(x)$  for all  $x \in \bar{X}$
  2. Reversing roles of  $v_1, v_2$  in (1)  $\Rightarrow v_2(x) \leq v_1(x)$  for all  $x \in \bar{X}$
  3. (1) and (2) imply  $v_1(x) = v_2(x)$  for all  $x \in \bar{X}$ , i.e. uniqueness.  $\square$

## Proof Sketch of Comparison Thm with smooth $v_1, v_2$

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- **Thm:**  $v_1 =$  subsolution,  $v_2 =$  supersolution  $\Rightarrow v_1(x) \leq v_2(x)$  all  $x$
- **Proof** by contradiction: suppose instead  $v_1(x) > v_2(x)$  for some  $x$
- ... equivalently,  $v_1 - v_2$  attains local maximum at some point  $x^* \in \bar{X}$  with  $v_1(x^*) > v_2(x^*)$
- Two cases:
  1.  $x^*$  in interior:  $x^* \in X$
  2.  $x^*$  on boundary:  $x^* = x_{\min}$  or  $x^* = x_{\max}$
- Here: ignore case 2, i.e. possibility of  $x^*$  on boundary
  - requires more work, just like non-smooth  $v_1, v_2$
  - for complete proof, see references

## Proof Sketch of Comparison Thm with smooth $v_1, v_2$

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- If  $v_1 - v_2$  attains local maximum at interior  $x^*$ , then since  $v_1, v_2$  are smooth also  $v_1'(x^*) = v_2'(x^*)$
- $v_1$  is subsolution means: for any smooth  $\phi$  and  $x^*$  such that  $v_1 - \phi$  attains local max:  $\rho v_1(x^*) \leq \max_{\alpha \in A} \{r(x^*, \alpha) + \phi'(x^*)f(x^*, \alpha)\}$
- In particular, use  $\phi = v_2$ :

$$\rho v_1(x^*) \leq \max_{\alpha \in A} \{r(x^*, \alpha) + v_2'(x^*)f(x^*, \alpha)\} \quad (1)$$

- Repeat symmetric steps with supersolution condition for  $v_2$ , setting  $\phi = v_1$  and noting that  $v_2 - v_1$  attains local min at  $x^*$

$$\rho v_2(x^*) \geq \max_{\alpha \in A} \{r(x^*, \alpha) + v_1'(x^*)f(x^*, \alpha)\} \quad (2)$$

- Subtracting (2) from (1) and recalling  $v_1'(x^*) = v_2'(x^*)$  we have  $v_1(x^*) \leq v_2(x^*)$ . Contradiction.  $\square$



## Proof Sketch of Comparison Thm, **non-smooth** $v_1, v_2$

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- Challenge: can no longer use  $\phi = v_2$  in subsolution condition because not differentiable (and similarly for  $\phi = v_1$ )
- Use **clever trick** to overcome this: use **two-dimensional** extension  $v_2(y) + \frac{1}{2\varepsilon} |x - y|^2$ 
  - key:  $v_2(y) + \frac{1}{2\varepsilon} |x - y|^2$  is smooth as function of  $x$
  - name of trick: **“doubling of variables”**
- For complete proof in **state-constrained case**, see
  - Soner (1986), Theorem 2.2
  - Capuzzo-Dolcetta and Lions (1990), Theorem III.1
  - Bardi and Capuzzo-Dolcetta (2008), Theorem IV.5.8

# Variants on Comparison Theorem (still $\Rightarrow$ Uniqueness)

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1. Known values of  $v$  at  $x_{\min}, x_{\max}$  (“Dirichlet boundary conditions”)
  - simplest case and the one covered in most textbooks, notes
  - kind of uninteresting, mostly useful for understanding proof strategy
  - Yu (2011), Section 3
  - Bressan (2011), Theorem 5.1
2. **Unbounded domain**  $x \in \mathbb{R}$ 
  - just like in discrete time, need certain boundedness assumptions on  $v_1, v_2$  (e.g. at most linear growth)
  - Sections 5 and 6 of Crandall (1995)
  - Theorem III.2.12 in Bardi & Capuzzo-Dolcetta (2008)

# Additional Results on HJB Equation in Economics

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- Strulovici and Szydlowski (2015) “On the smoothness of value functions and the existence of optimal strategies in diffusion models”
  - very nice analysis of **one-dimensional** case (note: does not apply to  $N > 1$ )
  - provide conditions under which  $v$  is **twice differentiable...**
  - ... in which case whole viscosity apparatus not needed
  - no uniqueness result...
  - ... but they note that uniqueness implied because  $v$ =solution to “sequence problem” which has unique solution