

# Optimal Timing of Decisions: A General Theory Based on Continuation Values

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# A Heuristic Example

## Example (Job Search I)

- 1 At each  $t$ , an unemployed worker obtains a wage offer  $w_t = w(Z_t)$ .
  - $(Z_t)_{t \geq 0}$ : the underlying state process.
- 2 Two choices:
  - accept the offer: work permanently at  $w_t$ .
  - reject the offer: receive compensation  $\tilde{c}_0$  and reconsider next period.
- 3 **Objective:** A stopping rule maximizing expected total returns.

- ① Value function based method
- ② Continuation value based method
  - Potential advantages
- ③ The theory we develop: continuation value based
  - Optimality results
  - Properties of continuation values and optimal policies

- $(Z_n)_{n \geq 0}$ : a time homogeneous Markov process

$$Z_n: (\Omega, \mathcal{F}, \{\mathcal{F}_n\}_{n \geq 0}, \mathbb{P}) \rightarrow (Z, \mathcal{Z})$$

- $P: Z \times \mathcal{Z} \rightarrow [0, 1]$  : the stochastic kernel of  $(Z_n)_{n \geq 0}$ 
  - $A \mapsto P(z, A)$  is a probability measure,  $\forall z \in Z$
  - $z \mapsto P(z, A)$  is  $\mathcal{Z}$ -measurable,  $\forall A \in \mathcal{Z}$
- $\mathcal{M}$ : the set of stopping times on  $\Omega$  w.r.t  $\{\mathcal{F}_n\}_{n \geq 0}$
- $r: Z \rightarrow \mathbb{R}$  : exit reward function
- $c: Z \rightarrow \mathbb{R}$  : flow continuation reward function
- $\beta \in (0, 1)$ : discount factor

# Value Function Based Method

The value function:

$$v^*(z) := \sup_{\tau \in \mathcal{M}} \mathbb{E}_z \left\{ \sum_{t=0}^{\tau-1} \beta^t c(Z_t) + \beta^\tau r(Z_\tau) \right\}.$$

The Bellman operator:

$$Tv(z) := \max \left\{ r(z), c(z) + \beta \int v(z') P(z, dz') \right\}.$$

Theorem (Stokey etc. (1989); Peskir and Shiryaev (2006))

If  $r, c \in bcZ$  and  $P$  is Feller, then  $v^*$  solves the Bellman equation:

$$v^*(z) = \max \left\{ r(z), c(z) + \beta \int v^*(z') P(z, dz') \right\},$$

and  $T: (bcZ, \|\cdot\|) \rightarrow (bcZ, \|\cdot\|)$  is a contraction mapping with unique fixed point  $v^*$ .

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# Value Function Based Method

## Example (Job search I cont.)

The exit and flow continuation rewards:

$$r(z) := \frac{u(w(z))}{1 - \beta} \quad \text{and} \quad c(z) \equiv c_0 := u(\tilde{c}_0).$$

The Bellman operator:

$$Tv(z) := \max \left\{ \frac{u(w(z))}{1 - \beta}, c_0 + \beta \int v(z')P(z, dz') \right\}.$$

If  $u$  is bounded and continuous and  $P$  is Feller, then

$$v^*(z) = \max \left\{ \frac{u(w(z))}{1 - \beta}, c_0 + \beta \int v^*(z')P(z, dz') \right\},$$

and  $T : bcZ \rightarrow bcZ$  is a contraction with unique fixed point  $v^*$ .



# Value Function Based Method: Limitations

## ① Unbounded rewards are common:

- AR(1) state process (unit root possible)

$$Z_{t+1} = b + \rho Z_t + \varepsilon_{t+1}, \quad (\varepsilon_t) \stackrel{\text{iid}}{\sim} N(0, \sigma^2), \quad \rho \in [-1, 1],$$

- Log-normal wage process:  $w(z) = e^z$
- CRRA utility

$$u(w) = \begin{cases} \frac{w^{1-\delta}}{1-\delta}, & \text{if } \delta \geq 0 \text{ and } \delta \neq 1 \\ \ln w, & \text{if } \delta = 1 \end{cases}$$

## ② Among others ...

# An Alternative Method

Recall the value function

$$v^*(z) = \max \left\{ r(z), c(z) + \beta \int v^*(z') P(z, dz') \right\}. \quad (1)$$

Define the continuation value function

$$\psi^*(z) := c(z) + \beta \int v^*(z') P(z, dz'). \quad (2)$$

Substituting (2) into (1):

$$v^*(z) = \max\{r(z), \psi^*(z)\}. \quad (3)$$

Substituting (3) into (2):

$$\psi^*(z) = c(z) + \beta \int \max\{r(z'), \psi^*(z')\} P(z, dz').$$

# Continuation Value Based Method

Define the continuation value operator, or Jovanovic operator

$$Q\psi(z) = c(z) + \beta \int \max \{r(z'), \psi(z')\} P(z, dz').$$

Example (Jovanovic, 1982)

Firm's decision: stay in or exit the industry?

$$V(x; p) = \pi(x; p) + \beta \int \max \{W, V(x'; p)\} P(x, dx').$$

- $V(x; p)$ : the expected value of staying in the industry
- $\pi(x; p)$ : the expected profit from the current industry (bounded)
- $W$ : expected return in a different industry (a constant)

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## Example (Job Search I cont.)

The Bellman operator

$$Tv(z) = \max \left\{ \frac{u(w(z))}{1 - \beta}, c_0 + \beta \int v(z') P(z, dz') \right\}.$$

The Jovanovic operator

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**Potential advantage–1:** smoother fixed points

- Facilitating analysis of the optimal policy.
- Facilitating analysis related to unbounded rewards.
- Computationally less expensive.

## Example (Job Search II)

A job search model with learning:

- The wage process

$$\ln w_t = \theta + \varepsilon_t, \quad (\varepsilon_t)_{t \geq 0} \stackrel{\text{iid}}{\sim} N(0, \gamma_\varepsilon)$$

- $\theta$ : unobservable component, prior belief  $N(\mu, \gamma)$
- If the offer is rejected, the agent observes  $w'$  and updates belief
- Posterior belief:  $\theta|w' \sim N(\mu', \gamma')$ 
  - $\gamma' = 1/(1/\gamma + 1/\gamma_\varepsilon)$  and  $\mu' = \gamma'(\mu/\gamma + \ln w'/\gamma_\varepsilon)$
- $f(w'|\mu, \gamma) = LN(\mu, \gamma + \gamma_\varepsilon)$ : the current expectation of the next period wage distribution.

## Example (Job Search II cont.)

The Bellman operator

$$Tv(w, \mu, \gamma) := \max \left\{ \frac{u(w)}{1 - \beta}, c_0 + \beta \int v(w', \mu', \gamma') f(w' | \mu, \gamma) dw' \right\}$$

and the Jovanovic operator

$$Q\psi(\mu, \gamma) := c_0 + \beta \int \max \left\{ \frac{u(w')}{1 - \beta}, \psi(\mu', \gamma') \right\} f(w' | \mu, \gamma) dw',$$

where  $f(w' | \mu, \gamma) = LN(\mu, \gamma + \gamma_\epsilon)$ .

**Note:**

- 3-dimensional ( $T$ )    v.s.    2-dimensional ( $Q$ ).



## **Potential advantage–2:** lower state dimension

- Simplifying challenging problems associated with
  - unbounded rewards
  - parametric continuity
  - differentiability
  - so on ...
  
- Mitigating the curse of dimensionality.

The first systematic study of optimal timing of decisions, based on continuation value functions and operators.

## Main results:

- A general optimality theory: bounded or unbounded rewards
- Conditions under which continuation values satisfy
  - (parametric) continuity
  - monotonicity
  - (continuous) differentiability
- Conditions under which threshold policies satisfy
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Individual applications of optimal timing:

- Jovanovic (1982), Posche (2010), Chatterjee and Rossi-Hansberg (2012), Kellogg (2014)

Unbounded dynamic programming:

- The weighted supremum norm theory
  - Boyd (1990), Alvarez and Stokey (1998), Le Van and Vailakis (2005)
- The local contraction theory
  - Rincon-Zapatero and Rodriguez-Palmero (2003), Martins-Da-Rocha and Vailakis (2010)

## Assumption 2.1

There exist a  $\mathcal{Z}$ -measurable function  $g : Z \rightarrow \mathbb{R}$  and constants  $n \in \mathbb{N}_0$  and  $a_1, \dots, a_4, m, d \in \mathbb{R}_+$  such that  $\beta m < 1$ , and, for all  $z \in Z$ ,

$$\int |r(z')| P^n(z, dz') \leq a_1 g(z) + a_2, \quad (4)$$

$$\int |c(z')| P^n(z, dz') \leq a_3 g(z) + a_4, \quad (5)$$

$$\text{and } \int g(z') P(z, dz') \leq m g(z) + d. \quad (6)$$

### Note:

- 1 If  $r$  and  $c$  are bounded, then assumption 2.1 holds.
- 2 True for some  $n \in \mathbb{N}_0 \implies$  true for all integer  $n' \geq n$ .
- 3 May use  $n_1$  in (4),  $n_2$  in (5), and  $n_1 \neq n_2$ .

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# Optimality Results

For  $\kappa : Z \rightarrow (0, +\infty)$ , the  $\kappa$ -weighted supremum norm:

$$\|f\|_{\kappa} := \sup_{z \in Z} \frac{|f(z)|}{\kappa(z)}.$$

Then  $(b_{\kappa}Z, \|\cdot\|_{\kappa})$  is a Banach space, where

$$b_{\kappa}Z := \{f \in m\mathcal{L} : \|f\|_{\kappa} < \infty\}.$$

Recall the Jovanovic operator:

$$Q\psi(z) := c(z) + \beta \int \max\{r(z'), \psi(z')\} P(z, dz').$$



## Theorem 2.1

Under assumption 2.1, there exist  $m', d' > 0$  such that for  $\ell : Z \rightarrow \mathbb{R}$  given by

$$\ell(z) := m' \left( \sum_{t=1}^{n-1} \mathbb{E}_z |r(Z_t)| + \sum_{t=0}^{n-1} \mathbb{E}_z |c(Z_t)| \right) + g(z) + d',$$

1.  $Q$  is a contraction mapping on  $(b_\ell Z, \|\cdot\|_\ell)$ .
2. The unique fixed point of  $Q$  in  $b_\ell Z$  is  $\psi^*$ .
3.  $\sigma^*(z) = \mathbb{1}\{r(z) \geq \psi^*(z)\}$  is an optimal policy.

### Note:

- 1 Assumption 2.1 holds for  $n = 0 \implies \ell(z) = g(z) + d'$ .
- 2 Assumption 2.1 holds for  $n = 1 \implies \ell(z) = m'|c(z)| + g(z) + d'$ .

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## Example (Job Search I cont.)

Let  $w(z) = e^z$ , and the state process

$$Z_{t+1} = b + \rho Z_t + \varepsilon_{t+1}, \quad (\varepsilon_t) \stackrel{\text{iid}}{\sim} N(0, \sigma^2), \quad \rho \in [-1, 1].$$

The agent's preference

$$u(w) = \begin{cases} \frac{w^{1-\delta}}{1-\delta}, & \text{if } \delta \geq 0 \text{ and } \delta \neq 1 \\ \ln w, & \text{if } \delta = 1 \end{cases}$$

The Jovanovic operator

$$Q\psi(z) = c_0 + \beta \int \max \left\{ \frac{u(w(z'))}{1-\beta}, \psi(z') \right\} f(z'|z) dz'.$$

- $f(z'|z) = N(\rho z + b, \sigma^2)$ .

## Example (Job Search I cont.)

Consider, e.g.,  $\delta \geq 0, \delta \neq 1$  and  $\rho \in [0, 1)$ .

- $r(z) = e^{(1-\delta)z} / ((1-\beta)(1-\delta))$  and  $c(z) \equiv c_0$ .

**Step 1.** Since  $X \sim N(\mu, \sigma^2) \implies \mathbb{E} e^{sX} = e^{s\mu + s^2\sigma^2/2}$  (MGF):

$$\int e^{(1-\delta)z'} P^t(z, dz') = b_t \underline{e^{\rho^t(1-\delta)z}} \quad (b_t \text{ is constant for fixed } t).$$

**Step 2.** Let  $\underline{g(z) = e^{\rho^n(1-\delta)z}}$  and apply MGF:

$$\int g(z') P(z, dz') \leq (g(z) + 1) \underline{e^{\rho^n \xi}} \quad (\xi \text{ is a constant}).$$

**Step 3.** Choose  $n \in \mathbb{N}_0$  s.t.  $\beta e^{\rho^n \xi} < 1$ , and let  $\underline{m = d = e^{\rho^n \xi}}$ .

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## Example (Job Search I cont.)

- The cases  $\rho \in (-1, 0]$ ,  $|\rho| = 1$  and  $\delta = 1$  can be treated similarly.

### Note:

- The local contraction method fails in this case.
  - Unbounded shocks and growth rate of the state process.

### Advantages of assumption 2.1:

- No further restrictions on the key parameters.
- Exploits the smoothing effect of future transitions.



## Example (Job Search II cont.)

The wage process

$$\ln w_t = \theta + \varepsilon_t, \quad (\varepsilon_t)_{t \geq 0} \stackrel{\text{iid}}{\sim} N(0, \gamma_\varepsilon).$$

The Jovanovic operator

$$Q\psi(\mu, \gamma) := c_0 + \beta \int \max \left\{ \frac{u(w')}{1 - \beta}, \psi(\mu', \gamma') \right\} f(w' | \mu, \gamma) dw'$$

- $\gamma' = 1 / (1/\gamma + 1/\gamma_\varepsilon)$  and  $\mu' = \gamma' (\mu/\gamma + \ln w' / \gamma_\varepsilon)$
- $f(w' | \mu, \gamma) = N(\mu, \gamma + \gamma_\varepsilon)$
- CRRA preference

$$u(w) = \begin{cases} \frac{w^{1-\delta}}{1-\delta}, & \text{if } \delta \geq 0 \text{ and } \delta \neq 1 \\ \ln w, & \text{if } \delta = 1 \end{cases}$$

## Example (Job Search II cont.)

Consider, e.g.,  $\delta = 1$ .

- $r(w) = \ln w / (1 - \beta)$  and  $c \equiv c_0$ .

**Step 1.** Use  $|\ln w| \leq w + 1/w$  and MGF:

$$\int |\ln w'| f(w' | \mu, \gamma) dw' \leq e^{\gamma \epsilon / 2} \left( e^{\mu + \gamma / 2} + e^{-\mu + \gamma / 2} \right).$$

**Step 2.** Let  $g(\mu, \gamma) = e^{\mu + \gamma / 2} + e^{-\mu + \gamma / 2}$ , and use MGF:

$$\int g(\mu', \gamma') f(w' | \mu, \gamma) dw' = g(\mu, \gamma).$$

**Step 3.** Let  $n = 1$ ,  $m = 1$  and  $d = 0$ .

## Example (Job Search II cont.)

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**Step 3.** Let  $n = 1$ ,  $m = 1$  and  $d = 0$ .

# Properties of Continuation Values: Continuity

## Assumption 3.1

- (1) The stochastic kernel  $P$  is Feller,
- (2)  $c, r, \ell$ , and  $z \mapsto \int |r(z')|P(z, dz')$ ,  $\int \ell(z')P(z, dz')$  are continuous.

Recall:

- Assumption 2.1 holds for  $n = 0 \implies \ell(z) = g(z) + d'$ .
- Assumption 2.1 holds for  $n = 1 \implies \ell(z) = m'|c(z)| + g(z) + d'$ .

## Proposition 3.1

If assumptions 2.1 and 3.1 hold, then  $\psi^*$  is continuous.

# Properties of Continuation Values: Differentiability

## Set up:

- $Z = Z^1 \times \dots \times Z^m \subset \mathbb{R}^m$
- $P$  has a density representation  $f$ :

$$P(z, A) = \int_A f(z'|z) dz' \quad \text{for all } A \in \mathcal{L}.$$

## Notations:

- $z = (z^1, \dots, z^m) \in Z$
- $z^{-i} = (z^1, \dots, z^{i-1}, z^{i+1}, \dots, z^m)$
- $D_i^j h(z) := \frac{\partial^j h(z)}{\partial (z^i)^j}$  and  $D_i^j f(z'|z) := \frac{\partial^j f(z'|z)}{\partial (z^i)^j}$

## Assumption 3.3

$D_i c(z)$  exists for all  $z \in \text{int}(Z)$  and  $i = 1, \dots, m$ .

# Properties of Continuation Values: Differentiability

## Assumption 3.4

$P$  has a density representation  $f$ , and for  $i = 1, \dots, m$ :

- (1)  $D_i^2 f(z'|z)$  exists for all  $(z, z') \in \text{int}(Z) \times Z$ ;
- (2)  $(z, z') \mapsto D_i f(z'|z)$  is continuous;
- (3) There are finite solutions of  $z^i$  to  $D_i^2 f(z'|z) = 0$  (denoted by  $z_i^*(z', z^{-i})$ ), and, for all  $z_0 \in \text{int}(Z)$ , there exist  $\delta > 0$  and a compact subset  $A \subset Z$  such that  $z' \notin A$  implies  $z_i^*(z', z_0^{-i}) \notin B_\delta(z_0^i)$ .

## Example (Job Search I cont.)

We show that assumption 3.4 holds:

- $P(z, A) = \int_A f(z'|z) dz'$ , where  $f(z'|z) = N(\rho z + b, \sigma^2)$ .
- $\frac{\partial^2 f(z'|z)}{\partial z^2} = 0$  has two solutions:  $z^*(z') = \frac{z' - b \pm \sigma}{\rho}$ .
- $|z'| \rightarrow \infty \implies |z^*(z')| \rightarrow \infty$ .



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- $|z'| \rightarrow \infty \implies |z^*(z')| \rightarrow \infty$ .

## Assumption 3.5

$k$  is continuous and  $\int |k(z')D_i f(z'|z)| dz' < \infty$  for all  $z \in \text{int}(Z)$ ,  $k \in \{r, \ell\}$  and  $i = 1, \dots, m$ .

## Proposition 3.3 (Differentiability)

Under assumptions 2.1 and 3.3–3.5,  $\psi^*$  is differentiable at interior points, and, for all  $z \in \text{int}(Z)$  and  $i = 1, \dots, m$ ,

$$D_i \psi^*(z) = D_i c(z) + \int \max \{r(z'), \psi(z')\} D_i f(z'|z) dz'.$$

## Assumption 3.6

For  $i = 1, \dots, m$ , the following conditions hold:

- (1)  $z \mapsto D_i c(z)$  is continuous on  $\text{int}(Z)$ ,
- (2)  $k$  and  $z \mapsto \int |k(z') D_i f(z'|z)| dz'$  are continuous on  $\text{int}(Z)$  for  $k \in \{r, \ell\}$ .

## Proposition 3.4 (Continuous Differentiability)

If assumptions 2.1, 3.4 and 3.6 hold, then  $z \mapsto D_i \psi^*(z)$  is continuous on  $\text{int}(Z)$  for  $i = 1, \dots, m$ .

# Properties of Continuation Values: Differentiability

Recall that

$$\ell(z) := m' \left( \sum_{t=1}^{n-1} \mathbb{E}_z |r(Z_t)| + \sum_{t=0}^{n-1} \mathbb{E}_z |c(Z_t)| \right) + g(z) + d'.$$

## Example (Job Search I cont.)

If  $\delta \geq 0$  and  $\delta \neq 1$ , then  $\mathbb{E}_z |r(Z_t)| = a_t e^{\rho^t(1-\delta)z}$  for some  $a_t > 0$ ,  $\forall t \geq 0$ .

- $f(z'|z) = N(\rho z + b, \sigma^2)$

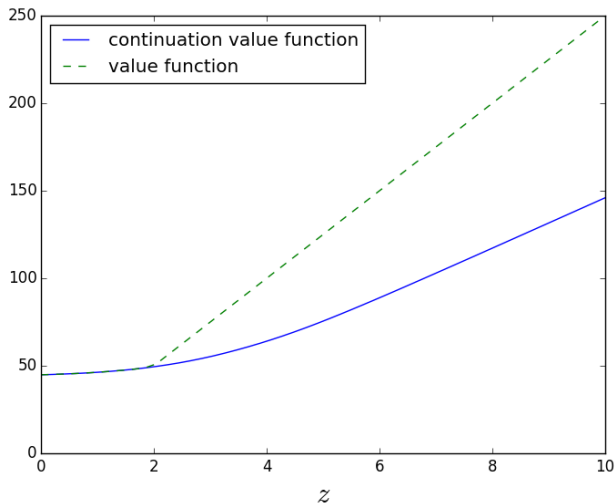
$$\implies z \mapsto \int e^{az'} \left| \frac{\partial f(z'|z)}{\partial z} \right| dz' \text{ is continuous for all } a \in \mathbb{R}.$$

$$\implies z \mapsto \int \left| r(z') \frac{\partial f(z'|z)}{\partial z} \right| dz', \int \left| \ell(z') \frac{\partial f(z'|z)}{\partial z} \right| dz' \text{ are continuous.}$$

Hence, assumption 3.6 holds, and  $\psi^*$  is continuously differentiable.

# Differentiability: VF v.s. CVF

Simulation:  $\beta = 0.96$ ,  $\rho = 0.6$ ,  $\delta = 1$ ,  $b = 0$  and  $c = 1$ .



# Optimal Policy

## Set up:

- $Z \subset \mathbb{R}^m$  with  $Z = X \times Y \subset \mathbb{R}^{m_0} \times \mathbb{R}^{m-m_0}$
- $(Z_t)_{t \geq 0} = \{(X_t, Y_t)\}_{t \geq 0}$
- Conditional independence: given  $Y_t$ , the next period states  $(X_{t+1}, Y_{t+1})$  are independent of  $X_t$ .

Then the stochastic kernel

$$P(z, dz') = P((x, y), d(x', y')) = dF_y(x', y').$$

- The flow continuation reward  $c : Y \rightarrow \mathbb{R}$ .

Then the Jovanovic operator

$$Q\psi(y) := c(y) + \beta \int \max\{r(x', y'), \psi(y')\} dF_y(x', y').$$

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In the next, we consider  $m_0 = 1$ .

## Assumption 4.1

$r$  is strictly monotone on  $X$ . Moreover, for all  $y \in Y$ , there exists  $x \in X$  such that  $r(x, y) = c(y) + \beta \int v^*(x', y') d\mathbb{F}_y(x', y')$ .

- $X_t$ : the threshold state
- $Y_t$ : the environment

## The reservation rule property

Under assumption 4.1, there is a decision threshold  $\bar{x} : Y \rightarrow X$ .

- When  $x$  attains  $\bar{x}$ , the agent is indifferent between stopping and continuing:  $r(\bar{x}(y), y) = \psi^*(y)$ , for all  $y \in Y$ .

## Example (Job Search II cont.)

The Bellman operator

$$Tv(w, \mu, \gamma) = \max \left\{ \frac{u(w)}{1-\beta}, c_0 + \beta \int v(w', \mu', \gamma') f(w'|\mu, \gamma) dw' \right\}.$$

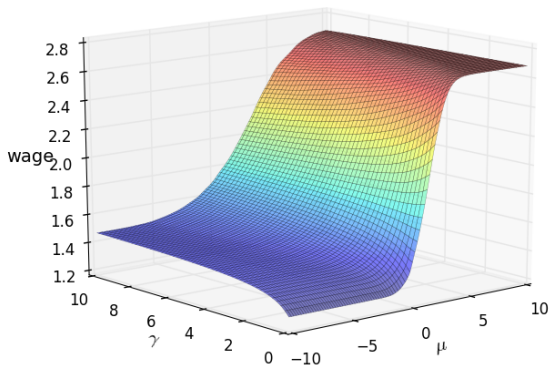
The Jovanovic operator

$$Q\psi(\mu, \gamma) = c_0 + \beta \int \max \left\{ \frac{u(w')}{1-\beta}, \psi(\mu', \gamma') \right\} f(w'|\mu, \gamma) dw'.$$

- $w =: x \in X := \mathbb{R}_{++}$ : threshold state
- $(\mu, \gamma) =: y \in Y := \mathbb{R} \times \mathbb{R}_{++}$ : environment
- $\bar{w} : Y \rightarrow X$ : the reservation wage

# Optimal Policy: The Reservation Wage

Simulation:  $\beta = 0.95$ ,  $\gamma_\varepsilon = 1.0$ ,  $\tilde{c}_0 = 0.6$ ,  $\delta = 3.0$



**CVI (2-dim)    v.s.    VFI (3-dim):**

**178 seconds    v.s.    More than 7 days!**

## Proposition 4.3 (Differentiability of Decision Threshold)

Let assumptions 2.1, 3.4, 3.6 and 4.1 hold. If  $r$  is continuously differentiable on  $\text{int}(Z)$ , then  $\bar{x}$  is continuously differentiable on  $\text{int}(Y)$ , with

$$D_i \bar{x}(y) = - \frac{D_i r(\bar{x}(y), y) - D_i \psi^*(y)}{D_x r(\bar{x}(y), y)} \text{ for all } y \in \text{int}(Y) \text{ and } i = 1, \dots, m.$$

- $r(x, y) - \psi^*(y)$ : terminating premium
- $D_i r(\bar{x}(y), y) - D_i \psi^*(y)$ : the marginal premium of  $y^i$
- $D_x r(\bar{x}(y), y)$ : the marginal premium of  $x$

## Extension–1: Repeated Sequential Decisions

- At each time  $t$ , the agent is either active or passive
- Active state: observe  $Z_t$ , continue or exit?
  - continuation:  $c(Z_t)$ , remain active at  $t + 1$
  - exit:  $s(Z_t)$ , transition to passive at time  $t$ , return to active with probability  $\alpha$  at time  $t + 1$
- E.g., Arellano (2008)

The value function:

$$v_a^*(z) = \max \left\{ v_p^*(z), c(z) + \beta \int v_a^*(z') P(z, dz') \right\}$$

$$v_p^*(z) = s(z) + \alpha \beta \int v_a^*(z') P(z, dz') + (1 - \alpha) \beta \int v_p^*(z') P(z, dz')$$

## Extension–2: Sequential Decision with More Choices

- At each period  $t$ , the agent observes  $Z_t$
- Choosing among  $N$  alternatives.
  - Alternative  $i$ : current reward  $r_i(Z_t)$ , stochastic kernel  $P_i(z, dz')$ .
- E.g., Jovanovic (1987), Moscarini and Postel-Vinay (2013)

The value function

$$v^*(z) = \max\{\psi_1^*(z), \dots, \psi_N^*(z)\}$$

$$\psi_i^*(z) = r_i(z) + \beta \int v^*(z') P_i(z, dz')$$

- Explore hidden advantages of the continuation value based method.
  - Smoothing effect of the future transitions
  - Conditional independence along the transition path
- Develop a general theory of optimal timing of decisions.
- Extend and improve the existing dynamic programming theory of optimal timing of decisions.
  - Analytically and computationally.



Thank you!