


Q-1(a)

If your current draw is same as discounted future expected draw, then you are indifferent. Also, if current draw is strictly better than next period's draw, you stop. You only redraw if your expected value from next draw is higher than current draw. This justifies structure of max operator.

Q-1(b)

This summer I was searching for place to work. Clearly, as I searched more I

could find a better, cheaper place.

But eventually I had to stop. Based on my belief about market & distribution of houses on the market, I made choice after searching for one and half month. This was my optimal stopping problem.

(C)

Take f & g two bounded continuous functions over $[0, 1]$.

Suppose $f(n) \geq g(n)$, then

$$(\beta f)(n) = \max \{x, \beta E f(x')\}$$

$$(\beta g)(n) = \max \{x, \beta E g(x')\}$$

Since, $E S(x') \leq E(f(x'))$

$$(\beta f)(x) \geq (\beta S)(x) \quad \forall n$$

β is monotone.

$$[\beta(f+a)](n) = \max\{x, \beta E(f+a)(x)\}$$

$$= \max\{x, \beta E(f(x)) + \beta a\}$$

obviously,

$$\max\{x, \beta E(f(x'))\} \leq \max\{x, \beta E(f(x)) + \beta a\}$$

for $\beta > 0, a \geq 0$. Hence, discount factor β exists. So, this Bellman operator satisfies Blackwell sufficient conditions and

hence is contraction map.

(d)

Contraction mapping property implies

B^w converges to unique fixed point of Bellman operator which will be our value function.

(e)

$$(B^w)(x) = \max_1 \{ x, \beta E^w(x) \}$$

$$E^w(x) = \int_0^1 x d\pi = 1$$

$$(B^w)(x) = \max \{ x, \beta \}$$

$$(\beta^2 w)(x) = \max \left\{ x, \beta \in \max \{x, \beta\} \right\}$$

$$\mathbb{E} \max \{x, \beta\} = \int_0^\beta \beta dx + \int_\beta^1 x dx$$

$$= \beta^2 + \left[\frac{1}{2} - \frac{\beta^2}{2} \right]$$

$$= \frac{\beta^2}{2} + \frac{1}{2}$$

$$(\beta^2 w)(x) = \max \left\{ x, \beta \left(\frac{\beta^2 + 1}{2} \right) \right\}$$

let us refer to fixed terms inside

fixed operator Cn. So,

$$c_1 = \beta \quad \text{and} \quad (\beta w)(x) = \max \{x, \beta\}$$

$$c_2 = \beta \left(\frac{\beta^2}{2} + \frac{1}{2} \right)$$

$$(\beta^3 \omega)(x) = \max_{\zeta_2} \left\{ x, \beta E \max \left\{ x, \zeta_2 \right\} \right\}$$

$$E \max \left\{ x, \zeta_2 \right\} = \int_{0}^{\zeta_2} c_2 dx + \int_{\zeta_2}^x \zeta'_2 dx'$$

$$= \frac{1}{2} + \frac{\zeta_2^2}{2}$$

$$(\beta^3 \omega)(x) = \max \left\{ x, \beta \left(\frac{1}{2} + \frac{\zeta_2^2}{2} \right) \right\}$$

Defn is $c_3 = \beta \left(\frac{1}{2} + \frac{\zeta_2^2}{2} \right)$

thus we will have

$$(\beta^n \omega)(x) = \max \left\{ x, c_n \right\}$$

where

$$c_0 = 1, \quad c_1 = \beta, \quad c_2 = \beta \left(\frac{1 + \beta^2}{2} \right)$$

$$c_3 = \beta \left(\frac{1 + c_2^2}{2} \right)$$

therefore,

$$c_n = \beta \left(\frac{1 + c_{n-1}^2}{2} \right), \quad c_0 = 1$$

$$\Rightarrow (B^n w)(x) = \max \left\{ x, c_n \right\}$$

Now we need to show that say
 c_n is convergent.

$$c_0 = 1, \quad c_1 = \beta, \quad c_2 = \beta \left(1 + \beta^2 \right)$$

We know $c_2 \leq \beta$. Suppose

$$c_{n-1} \leq \beta$$

$$c_{n-1}^2 \leq \beta^2 \leq 1$$

$$1 + c_{n-1}^2 \leq 1$$

$$\frac{1 + c_{n-1}^2}{2} \leq \frac{1}{2}$$

$$\left(\frac{1 + c_{n-1}^2}{2} \right) \beta \leq \frac{\beta}{2} \leq \beta$$

$$\Rightarrow c_n \leq \beta \quad \forall n$$

so seqn c_n is bounded in \mathbb{R} and

by completeness in \mathbb{R} , it has limit.

Suppose $c_n \xrightarrow{*} \hat{c}$

$$c_n = \frac{\beta}{2} (1 + c_{n-1}^2)$$

$$c_* = \frac{\beta}{2} (1 + c_*^L)$$

$$2c^* = \beta + \beta c^{*^2}$$

$$\beta c^{*^2} - 2c^* + \beta = 0$$

$$c^* = \frac{1}{\beta} \left(1 \pm \sqrt{1 - \beta^2} \right)$$

Since $c_n > 0 \quad \forall n, \quad c^* > 0 \quad \text{So}$

$$c^* = \frac{1}{\beta} \left(1 - \sqrt{1 - \beta^2} \right)$$

using $\rho = -\log(\beta) \Rightarrow \beta = e^{-\rho}$

$$c^* = e^\rho \left(1 - \left(1 - e^{-\rho} \right)^{\frac{1}{2}} \right)$$

Hence
 $(B^n w)(x) \rightarrow \max \{x, c^*\} = v(x)$

$$V(x) = \begin{cases} x^* & \text{if } x \leq n^* \\ x & \text{if } x \geq n^* \end{cases}$$

$$\text{where } x^* = e^f \left(1 - \left(1 - e^{-2f} \right)^{1/2} \right)$$



Q-2

Once you set draw of cost, you can choose to work on project and pay cost c , then you expect pay-off in terms of cost is

$$\underbrace{(1-p)c}_{\text{complete}} + \underbrace{p(c+l+EV(c'))}_{c, l \text{ cost r now draw } c'}$$

If you don't work on it, you just get $EV(c')$ but pay late fee l .

$$V(c) = \min \left\{ (1-p)c + p(c+l+EV(c')), l+EV(c') \right\}$$

If minimum base problem is set in terms of cost,

$$V(c) = \min \left\{ c + p(l + EV(c')), l + EV(V(c')) \right\}$$

To match it with our usual specification,

$$\min \left\{ c + p(l + EV(c')), l + EV(V(c')) \right\} = V(c)$$

$$= \min \left\{ c, (1-p)(l + EV(c')) \right\}$$

Since there is late fee and you are incurring cost, it is OK to not assume discounting. We can say $(1-p)$ is discounting as you pay l with prob $1-p$.

Assuming fixed point exists,

$$V(c) = \min \left\{ c + p(l + EV(c')), l + EV(c') \right\}$$

At optimum,

$$c + p(l + EV(c')) = l + EV(c')$$

Now,

$$\begin{aligned} EV(c') &= \int_0^c (c + pl + pEV(c')) dc \\ &\quad + \int_c^1 (l + EV(c')) dc \\ &= pl + \frac{c^2}{2} + pEV(c') + l + EV(c') \\ &\quad - (l + EV(c')) c \end{aligned}$$

$$0 = (p+1)l + \frac{c^2}{2} + (p-c)EV(c') - l c$$

$$V(c) = \min \left\{ c, \beta(l + V(c')) \right\}$$

$$\hat{c}^* = l + \int_0^1 V(c') d c$$

$$= l + \beta \int_0^{\hat{c}} c + \beta \int_{\hat{c}}^1 c' d c$$

$$= \beta \left\{ l + \frac{c^{*2}}{2} + c^* - \hat{c}^2 \right\}$$

$$c^* = \beta \left\{ l + c^* - \frac{c^{*2}}{2} \right\}$$

$$\frac{\beta}{2} c^{*2} + (l - \beta) c^* - \beta l = 0$$

$$C^* = \frac{(\beta - 1) \pm \sqrt{[(1-\beta)^2 - 4(\beta/2)(-\beta\ell)]}}{2\beta/2}$$

$$C^* = \frac{(\beta - 1) \pm \sqrt{(1-\beta)^2 + 2\beta^2\ell}}{\beta}$$

$$C^* = \frac{1}{\beta} \left[\beta - 1 + \sqrt{(1-\beta)^2 + 2\beta^2\ell} \right]$$

becu $C \geq 0$.

(d)

Since you are weighting different periods now at weight β , ρ does matter as it determines how likely you are to succeed in given period.

(e)

$$P(\text{under take}) = P_r(C \leq c) = c^*$$

since c is uniform.

$$P_{\text{complete under take}} = 1 - P$$

$$P_{\text{success}} = P_{\text{under take}} \times P_{\text{complete success}} = c^*(1 - P)$$

This probability is constant each period as all draw are independent.

Then its geometric distribution with prob of success $c^*(1 - P)$.

Then mean is $\frac{1}{c^*(1 - P)}$.

Q-3 (a)

The flow budget constraint is,

$$\frac{d}{dt}(P_t a_t) = i_t (P_t a_t) - P_t c_t + P_t y_t$$

$$P_t \dot{a}_t + a_t \dot{P}_t = i_t P_t a_t - P_t c_t + P_t y_t$$

$$\dot{a}_t = i_t a_t - a_t \frac{\dot{P}_t}{P_t} - c_t + y_t$$

$$\dot{a}_t = (i_t - \bar{r}_t) a_t - \frac{P_t}{\dot{P}_t} c_t + y_t$$

$$\dot{a}_t - (i_t - \bar{r}_t) a_t = y_t - c_t$$

(b)

We can solve this differential equation
to get lifetime budget constraint.

Define real interest rate $\varrho(t) = i_t - \bar{r}_t$

$$\dot{a}_t - \delta(t) a_t = y_t - c_t$$

$$e^{-\int_0^t \delta(s) ds} \dot{a}_t - e^{-\int_0^t \delta(s) ds} \delta(t) a_t = (y_t - c_t) e^{-\int_0^t \delta(s) ds}$$

$$\frac{d}{dt} \left(e^{-\int_0^t \delta(s) ds} a_t \right) = e^{-\int_0^t \delta(s) ds} (y_t - c_t)$$

Taking lifetime integral,

$$\int_0^\infty \frac{d}{dt} \left(e^{-\int_0^t \delta(s) ds} a_t \right) dt = \int_0^\infty e^{-\int_0^t \delta(s) ds} (y_t - c_t) dt$$

$$\lim_{t \rightarrow \infty} e^{-\int_0^t \delta(s) ds} a_t - a_0 = \int_0^\infty e^{-\int_0^t \delta(s) ds} y_t - \int_0^\infty e^{-\int_0^t \delta(s) ds} c_t$$

It says present value of terminal wealth

minus current wealth plus present value
of total income is equal to present value
of lifetime consumption. Usually we

assume in models that $\lim_{t \rightarrow \infty} e^{\rho} a_t = 0$

as you don't end up with final wealth if
you optimize.

$$a_0 + \int_0^\infty e^{\rho s} c(s) ds = y_t = \int_0^t e^{\rho s} r(s) ds - c_t$$

Although I think the other part in

a_t should only drop out when we

optimize. Budget constraint itself is correct as solution to DE.

(c)

Now we can write budget constraint as

$$a_0 + \int_0^{\infty} e^{-\int_s^t r(s) ds} y_t dt = \int_0^{\infty} e^{-\int_s^t r(s) ds} c_t dt$$

$$\mathcal{L} = \int_0^{\infty} e^{-pt} u(c_t) dt + \lambda \left[\int_0^{\infty} e^{-\int_s^t r(s) ds} (y_t - c_t) dt + a_0 \right]$$

$$\frac{\partial \mathcal{L}}{\partial c_t} = e^{-pt} u'(c_t) - \lambda e^{-\int_s^t r(s) ds} = 0$$

$$\Rightarrow u'(c_t) = e^{\int_s^t r(s) ds} e^{pt}$$

Taking log & then hine derivative

$$\log(u'(ct)) = - \int_0^t r(s) ds + ft$$

Derivative

$$\frac{1}{u'(ct)} u''(ct) \dot{c}_t = -r(t) + f = p - r(t)$$

$$\frac{u''(ct)}{u'(ct)} \dot{c}_t = p - r(t)$$

(d)

$$u(ct) = \frac{c_t^{1-r}}{1-r}$$

$$u'(ct) = \bar{c}_t^{-r} - (r+1)$$

$$u''(ct) = -r \bar{c}_t$$

$$\frac{-\gamma c_t}{c_t^{-\gamma}} \cdot \dot{c}_t = p - r(t)$$

$$-\gamma c_t^{\gamma-1} \dot{c}_t = p - r(t)$$

$$\frac{\dot{c}_t}{c_t} = \frac{x(t) - p}{\gamma}$$

- Euler Equation
for CRRA

if $u(c_t) = \log(c_t)$

$$u'(c_t) = \frac{1}{c_t}$$

$$u''(c_t) = -\frac{1}{c_t^2}$$

$$\frac{-1}{c_t^2} \times c_t \cdot \dot{c}_t = p - r(t)$$

$$\frac{\dot{c}_t}{c_t} = p - r(t)$$

(e)

The optimal control problem is

$$\max_{\{c_t\}_{t \geq 0}} \int_0^{\infty} e^{-ft} u(c_t) dt$$

$$\text{s.t. } \dot{a}_t = \varphi(t) a_t = y_t - c_t$$

Control: c_t

State: $a_t, \varphi(t), y_t$

Multiplic: λ_t

(f)

Hamiltonian,

$$H(c_t, y_t, a_t, \varphi(t)) = u(c_t) + \lambda_t [y_t - c_t + \varphi(t) a_t]$$

(g)

$$0 = \frac{\partial H}{\partial c} = u'(c_t) - h_t = 0$$

$$\rho h_t - \dot{h}_t = \frac{\partial H}{\partial a_t} = h_t r(t) = 0$$

$$\dot{a}_t = y_t - c_t + \varphi(t)a_t$$

$$\Rightarrow u'(c_t) = h_t$$

Taking time derivative

$$u''(c_t) \dot{c}_t = \dot{h}_t = \rho h_t - h_t r(t)$$

$$u''(c_t) \dot{c}_t = \rho u'(c_t) - u'(c_t) r(t)$$

$$\frac{u''(c_t)}{u'(c_t)} \dot{c}_t = \rho - r(t)$$

This is same as Euler equation

Using simple Lagrangian with lifetime budget constraint.

Q-4

$$V_0 = \max_{\{c_t\}_{t>0}} \int_0^{\infty} e^{-\beta t} u(c_t)$$

$$\dot{a}_t = \alpha_t a_t + y_t - c_t$$

a_t is only state variable that matters for optimal choice because α_t, y_t are exogenous and agent's choice does not change their path. Moreover, their path is perfectly known they are like constants.

The value function depends time because there are exogenous processes y_t, α_t that are not state but time dependent process.

(b)

Discrete time Bellman Equation is,

$$V_t(a_t) = \max_c \left\{ u(c_t) \Delta t + \frac{1}{1 + p_{\Delta t}} V_{t+\Delta}(a_{t+\Delta}) \right\}$$

$$(1 + f_{\Delta t}) V_t(a_t) = \max_c \left\{ u(c_t) \Delta t + f u(c_t) \Delta t^2 + V_{t+\Delta}(a_{t+\Delta}) \right\}$$

$$f_{\Delta t} V_t(a_t) = \max_c \left\{ u(c_t) \Delta t + f u(c_t) \Delta t^2 + V_{t+\Delta}(a_{t+\Delta}) - V_t(c_t) \right\}$$

$$p_{\Delta t} V_t(a_t) = \max_{c_t} \left\{ u(c_t) \Delta t + p u(c_t) \Delta t^2 + V_{t+\Delta}(a_{t+\Delta}) - V_t(c_t) \right\}$$

Now we can take Taylor Series expansion

of $V_{t+\Delta t}(a_{t+\Delta t})$ around (a_t, t) .

$$V(a_{t+\Delta t}, t+\Delta t) = V(a_t, t) + \frac{\partial V}{\partial t} \Delta t + \frac{\partial V}{\partial a_t} (a_{t+\Delta t} - a_t)$$

then

$$PV_t(a_t) = \max_{c_t} \left\{ u(c_t) + \rho u(q) \Delta t + \frac{\partial V}{\partial t} + \frac{\partial V}{\partial a_t} \cdot \frac{(q_t + \Delta t - a_t)}{\Delta t} \right\}$$

Taking $\Delta t \rightarrow 0$

$$PV_t(a_t) = \max_{c_t} \left\{ u(c_t) + \frac{\partial V}{\partial t} + \frac{\partial V}{\partial a_t} \dot{a}_t \right\}$$

$$= \frac{\partial V}{\partial t} + \max_{c_t} \left\{ u(c_t) + \frac{\partial V}{\partial a_t} \left(r_t a_t + y_t - c_t \right) \right\}$$

We can take out $\frac{\partial V}{\partial t}$ because it does not depend on our choice c_t but only on a_t .

(c)

$V_t(a)$ term appears bcoz our value function not only depend on initial state but actual time when you start optimizing as path y_t, r_t depends on that.

(d)

$$u'(c) - \frac{\partial V_t(a)}{\partial a} = 0$$

This says marginal utility of current consumption is equal to lifetime value of investing that extra unit.

A consumption policy function is $c_t(a_t)$

that takes current state at t time

which gives $r_f r y_t$ and give optimal value of c_t in each time t .

Since u' is one-to-one function, we

can define

$$c(a) = (u')^{-1} \left\{ \frac{\partial v(a)}{\partial a} \right\}$$

which is in general non-linear PDE.

$$f v(a_t, t) = \frac{\partial v}{\partial t} + u \left((u')^{-1} \left(\frac{\partial v}{\partial a} \right) \right) + \frac{\partial v}{\partial a} \left[r_a + y_t - (u')^{-1} \left(\frac{\partial v}{\partial a} \right) \right]$$

This is non-linear PDE as function

$(u')^{-1}$ is non-linear.

(f)

$$\frac{\partial V_t(a)}{\partial a} = \frac{\partial V_t(a)}{\partial t \partial a} + \frac{\partial^2 V_t(a)}{\partial a^2} [x_t a + y_t - c] + l_t \frac{\partial V_t(a)}{\partial a}$$

(g)

$$\frac{d V(t, a_t)}{dt} = \frac{\partial V(t, a_t)}{\partial t} + \frac{\partial V}{\partial a} \times \frac{\partial a}{\partial t}$$

(h)

$$\frac{d V(t, a_t)}{dt} = \frac{\partial V(t, a_t)}{\partial t} + \frac{\partial V}{\partial a} \times \frac{\partial a}{\partial t}$$

$$U'(c) = \frac{\partial V(t, a)}{\partial a}$$

$$\frac{\partial V_t(a)}{\partial a} = \frac{\partial V_t(a)}{\partial t \partial a} + \frac{\partial^2 V_t(a)}{\partial a^2} [x_t a + y_t - c] + l_t \frac{\partial V_t(a)}{\partial a}$$

Taking first derivative of FOC,

$$u''(c) \dot{c} = \frac{\partial^2 V(t, a)}{\partial a \partial t} + \frac{\partial^2 V(t, a)}{\partial^2 a} \dot{a}$$

Now from HJB,

$$\frac{\partial V}{\partial t} = pV(t, a) - u(c) - \frac{\partial V}{\partial a} \dot{a}$$

$$\Rightarrow \frac{\partial V}{\partial t \partial a} = p \frac{\partial V}{\partial a} - \frac{\partial V^2}{\partial a^2} \dot{a} - \frac{\partial V}{\partial a} \lambda_t$$

$$\therefore u''(c) \dot{c} = p \frac{\partial V}{\partial a} - \frac{\partial V}{\partial a} \lambda_t$$

$$\Rightarrow u''(c) \dot{c} = (p - \gamma_t) u'(c)$$

$$\Rightarrow \frac{u''(c)c}{u'(c)} = f - r_f \quad \text{Euler Equation}$$



(Q-5)

At any time t , you have extracted total oil of $\int_0^t c_s ds$, so remaining oil at time t , $x_t = x_0 - \int_0^t c_s ds$

(b)

First we can find flow budget constraint,

$$i_t = -c_t$$

State: x_t

Control: c_t

Multiplicies: λ_t

then

$$H = u(c_t) + \lambda_t [-c_t]$$

(c)

$$\frac{\partial H}{\partial c} = 0 = u'(c) - \lambda = 0$$

$$f\lambda_t - \dot{\lambda}_t = \frac{\partial H}{\partial \lambda} = 0$$

$$f\lambda_t = \dot{\lambda}_t$$

$$\dot{c}_t = -\dot{\lambda}_t$$

using $\frac{\dot{\lambda}_t}{\lambda_t} = f$

$$\Rightarrow \lambda_t = \lambda_0 e^{ft}$$

$$\Rightarrow u'(c_t) = \lambda_0 e^{ft}$$

$$\Rightarrow \frac{1}{ct} = H_0 e^{ft}$$

$$\Rightarrow ct = \frac{1}{H_0} e^{-ft}$$

Now with budget constraint,

$$x_0 = \int_0^{\infty} \frac{1}{H_0} \cdot e^{-ft} dt$$

$$x_0 = \frac{1}{H_0} \times \left[\frac{-f t}{-P} \right]_0^\infty = \frac{1}{H_0} \times \frac{1}{P}$$

$$x_0 f = \frac{1}{H_0}$$

$$\Rightarrow ct = x_0 f e^{-ft}$$

(d) - (e)

HJB equation is,

$$fV(x) = \max_{c>0} \left\{ \ln(c) - c \frac{\partial V}{\partial x} \right\}$$

$$= \max_{c>0} \left\{ \ln(c) - c \cdot \frac{\partial V}{\partial x} \right\}$$

Now, $\frac{\partial V}{\partial x} = \frac{b}{x}$

$$f_a + f_b \ln(x_0) = \max_{c>0} \left\{ \ln(c) - c \cdot \frac{b}{x} \right\}$$

RHS is maximized at

$$\frac{1}{c} = \frac{b}{x}$$

$$c = x/b$$

$$f_a + f_b \ln(n_0) = \ln(x_b) - \frac{x}{b} \times \frac{b}{x}$$

$$= \ln(a) - \ln(b) - 1$$

$$\Rightarrow f_a = -\ln(b) - 1$$

$$f_b = 1$$

$$\Rightarrow b = 1/f$$

$$a = -\frac{\ln(1/f) - 1}{f} = \frac{\ln(f) - 1}{f}$$