

ECON 204B Cheat Sheet for Midterm

Decision Theory

Expected Utility Hypothesis

$$\max_{a \in A} E_{\pi} u = \max_{a \in A} \sum_{s \in S} \pi(s) u(x(a, s))$$

CRRA

$$u(x|r) = \frac{x^{1-r}}{1-r}$$

CARA

$$u(x|a) = 1 - e^{-ax}$$

Coefficient of Absolute Risk Aversion

$$A(x) = -\frac{u'(x)}{u''(x)}$$

Coefficient of Relative Risk Aversion

$$R(x) = -\frac{u''(x)}{u'(x)} x = xA(x)$$

Risk Measurement

$$Var_L[x] = E_L(x - E_L x)^2$$

$$= \sum_{i=1}^N p_i (x_i - \bar{x})^2$$

$$\bar{x} = E_L x = \sum_{i=1}^N p_i x_i$$

$$\sigma_L = \sqrt{Var_L}$$

Stochastic Dominance

$$F(x) \geq G(x) \quad \forall x \Rightarrow$$

$$\mu_F = \mu_G, \int_{-\infty}^x F(t) dt \geq \int_{-\infty}^x G(t) dt \forall x \Rightarrow$$

Definitions and Conditional Probabilities

$$p(s) = \sum_{z \in Z} p(s, z) \quad \text{Prior prob. of } s$$

$$p(z) = \sum_{s \in S} p(s, z) \quad \text{Prob. of message } z$$

$$p(z|s) = \frac{p(s, z)}{p(s)} \quad \text{Likelihood}$$

$$p(s|z) = \frac{p(s, z)}{p(z)} \quad \text{Posterior Probability}$$

Bayes Theorem

$$p(s|z) = \frac{p(z|s)p(s)}{p(z)}$$

$$p(s|z) = \frac{p(z|s)p(s)}{\sum_{t \in S} p(z|t)p(t)}$$

$$\frac{p(s|z)}{p(t|z)} = \frac{p(z|s)p(s)}{p(z|t)p(t)}$$

$$\ln\left(\frac{p(s|z)}{p(t|z)}\right) = \ln\left(\frac{p(z|s)}{p(z|t)}\right) + \ln\left(\frac{p(s)}{p(t)}\right)$$

Value of Information

$$VI = \sum_{z \in Z} p(z) \sum_{s \in S} p(s|z) [u_z^*(s) - u_0^*(s)]$$

$$\text{solves } 0 = \sum_{z \in Z} \sum_{s \in S} p(z) p(s|z) u(a_z^* - x, s)$$

$$x \in (-\infty, \infty)$$

Variance

Variance

Mean

Std Err.

G FOSDs F

G SOSDs F

(Risk Neutral)

(General Case)

Dynamic Programming

$$V(y_0, 0) = \max_{a_t \in A_t} \sum_{t=0}^T F(a_t, y_t, t)$$

$$\text{s.t } y_{t+1} = y_t + Q(a_t, y_t, t)$$

$$G(a_t, y_t, t) \geq 0$$

Game Theory Basics

Strategy s_i is *weakly dominant* if:

$$f_i(s_i, s_{-i}) \geq f_i(s'_i, s_{-i}) \quad \forall s'_i \in S_i, \quad s_{-i} \in S_{-i}$$

A pure strategy $s_i \in B_i(s_{-i})$ is a *best response* to profile s_{-i} if:

$$f_i(s_i, s_{-i}) \geq f_i(s'_i, s_{-i}) \quad \forall s'_i \in S_i$$

Nash Equilibrium

A profile $s_i^* = (s_1^*, \dots, s_n^*)$ is a Nash Equilibrium if:

$$s_i^* \in B(s_{-i}^*) \quad i = 1, \dots, n$$

Result: Normal Form Game Solution

Start with a NFG or EFG;

while *there exists a dominant strategy* **do**

 Eliminate strictly dominated strategies and reduce the game;

if *Only one profile remains* **then**

 | it is the DS solution;

end

end

if *Only one profile remains* **then**

 | it is the IDDS solution;

end

Inspect for mutual BR \rightarrow These are pure NE;

Check for mixed NE $\sigma_i \in B_i(\sigma_{-i})$;

Algorithm 1: Finding NFG Solutions

Payoff Function for Mixed NE

Player i 's expected payoff when other players' pure strategy is s_{-i} :

$$f_i(\sigma_i, s_{-i}) = \sum_{k=1}^N p_k f_i(t_k, s_{-i})$$

Player i 's expected payoff when playing own pure strategy t_i :

$$f_i(t_k, \sigma_{-i}) = \sum_{j=1}^m q_j f_i(t_k, s_{-i}^j)$$

This is used to find mixed NE. For example, in the 2×2 case:

$$f_1(s_1, \sigma_{-1}) = f_1(s_2, \sigma_{-1})$$

$$f_2(s_1, \sigma_{-2}) = f_2(s_2, \sigma_{-2})$$

Extensive Form Games

Procedure for finding solution of EFGs with perfect information

1. Convert penultimate nodes ν into terminal nodes

(a) If ν is owned by player i , then he chooses the maximum payoff

(b) If ν is owned by Nature, take expectation over the payoff vector

2. Iterate over step 1 until reaching the initial node
3. Reconstruct each player's strategy for their choices in steps 1-2
4. The resulting profile is subgame perfect Nash equilibrium (SPNE)
5. In the case if *imperfect information*, find the smallest subgame that contains terminal nodes. Find all NE of that subgame. Replace the initial node of that subgame by a NE payoff vector. Iterate to a solution and get one SPNE. Then loop on step 1 until all N equilibria in the minimal game have been used.

Bayesian Nash Equilibrium, PBE, and Seq. EQ

1. Beliefs μ_i at each info set for player i are consistent with *common prior* and likelihood from s_{-i}^* and own realized type $\bar{\theta}_i$ via Bayes
2. At each info set, player i maximizes $E(u_i | \mu_i)$:

$$E_{\theta_i} [u_i(s_i^*(\bar{\theta}_i), s_{-i}^*(\theta_{-i}), \theta_i) | \bar{\theta}_i] \geq E_{\theta_i} [u_i(s_i'(\bar{\theta}_i), s_{-i}^*(\theta_{-i}), \theta_i) | \bar{\theta}_i]$$

3. Previous items hold in every subgame
4. Robust to sufficiently small trembles

- 1 and 2 constitute a *Bayesian Nash Equilibrium*
- 1 - 3 constitute a *Perfect Bayesian NE*
- 1 - 4 constitute a *Sequential Equilibrium*