

EC117 Schedule

Jan 29 HW1

Feb 12 HW2.

Feb 24 Proposal

March 9 Exam 1

April 6 Draft due.

April 22 Exam 2.

April 27 Last class

May 7. Pre.

Week 1. Intro.

— 苏格拉底 opposed to writing (new tech)

Plato documented it.

— Social Media, politics / fake news manipulation
10 yrs ago, tech abused.

Web and internet (networks of computers)

↳ Berners-Lee: HTML.

• contract for the web.

regulating

EU: general data protection regulation

(cookie ...)

— Social elements in economic decisions.

Interactions direct agent \Leftrightarrow agent.

Influences do not go through market

(not reflected by prices)

Firms \Leftrightarrow Firms

- Duvaris \Leftrightarrow Brogen.

Government \Leftrightarrow G

- Local, state, national

e.g. Amazon headquarters

National: Treaties, War...

↳ COW MIDB dataset.

• Nato. (1950+) (war database)

• WARSAW PACT.

• Trade

— Casual inference, about phenomena.

• A special case: Polarization in D.C.

1990 votes vs 2015 votes

— Setting: Agent i , decision Y_i .

X_i , vector of characteristics.

Neighborhood: D_i

$$Y_i = F(X_i, \text{Reference Group};)$$

Individual effect.

Neighborhood: $\gamma(i)$

$X_{\gamma(i)}$: aggregates of characteristics.

Contextual effect

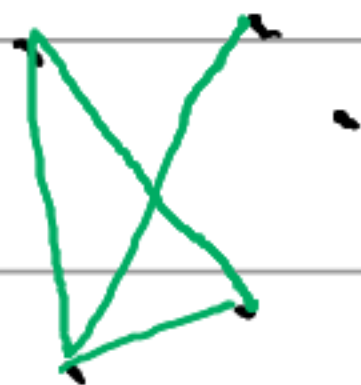
$\gamma(i) \rightarrow$ Endogenous social effect.

eg. $GPA_i = f(i's \text{ characteristics})$,

Neighborhoods are $\begin{cases} \text{exogenous.} \\ \text{endogenous.} \end{cases}$

- Social Networks

- Social graph.



- Algebraic representation.

Agent	a	b	c
a	0	1	0
b	1	0	1
c	0	1	0

Sociomatrix.

Adjacency matrix.

Week 2. 20/01/22.

- Precise Technology.

Social Interactions vs. Social Network.

$$Y_i = X_i, X_{\gamma(i)}, Y_{\gamma(i)}$$

\uparrow own \uparrow Neighbors community \uparrow Decisions of social contacts.

• QJE Sacerdote, Bruce (2001). Peer effect, Dartmouth

$$GPA_i = ACA_i, \underbrace{ACA_j}_{\text{other's academic standing}}, GPA_j, \text{other factors} \dots$$

$$\underbrace{GPA_{\gamma(i)}}_{\text{e.g. dorm's GPA (average)}}$$

Assumption: (i, j) communicate (GPA_i, GPA_j) are determined simultaneously.

$$GPA_i = \delta_i + \alpha ACA_i + \beta ACA_j + \gamma \cdot GPA_j$$

$$GPA_j = \delta_j + \beta ACA_i + \alpha ACA_j + \gamma GPA_i$$

$$GPA_i, GPA_j = F(A_{CA_i}, A_{CA_j})$$

Variation of theory: Decision_i depends on i's expectation of j's academic performance.

Y_i can be stuff like discrete variable ...
(Joining clubs ...)

- Reference Group: Exogenous / Endogenous.

• Endogeneity in Empirical Economics.
identification!

- Social Network: Decisions depend on specific economic agents.

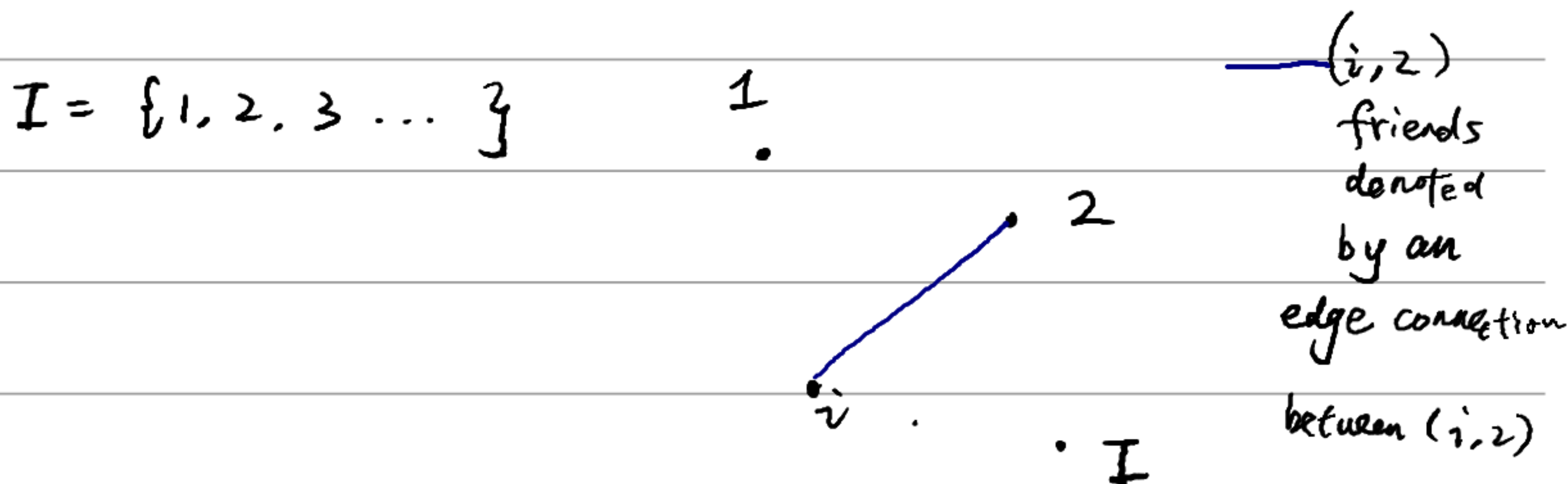
• Graphs, networks.

(Königsberg Bridge problem
Euler: 7 bridges.
↳ Eulerian Path)

• Diffusion of Microfinance in India. To expose villages to microfinance, they need a way

to get the word out of influencers.

(Castes .



$[A]: I \times I$ ① $a_{ij} = 1$ if (i, j) friends communicate.
0 otherwise.

{ Social Matrix
Adjacent Matrix . ② a_{ij} : weight, not necessarily $\{0, 1\}$.
weighted graph.

③ unsymmetric (like follower ..)

i follow $j \Rightarrow a_{ij} = 1, a_{ji} = 0.$

e.g Adhealth dataset.

\Rightarrow

	1	2	3	4
1	0	0	0	0
2	1	0	0	0
3	0	1	0	0
4	0	0	1	0

Looking up / following

Someone else is

a choose.

Graphs not symmetric.

eg. Web graph is fundamentally a directional and weighted graph.

$$Y_i = (X_i, X_{\delta(i)}, Y_{\delta(i)}) \quad \text{or } E(Y_{\delta(i)})$$

↳ Classical interaction setting

Social Network, we specify influences to (from) specific other individual.

Week 3 20/01/27

— Sacredote (2007): Dartmouth, randomly assigned roommates, social multiplier.
(policy).

- Implication: How to group students? By ability or...

- Empirical: Dartmouth 1997-1998 #2000 ~ #1589

Pretreatment: Parallel assumption

③ 1.7 + mailman $\frac{3}{4}$ + evil.

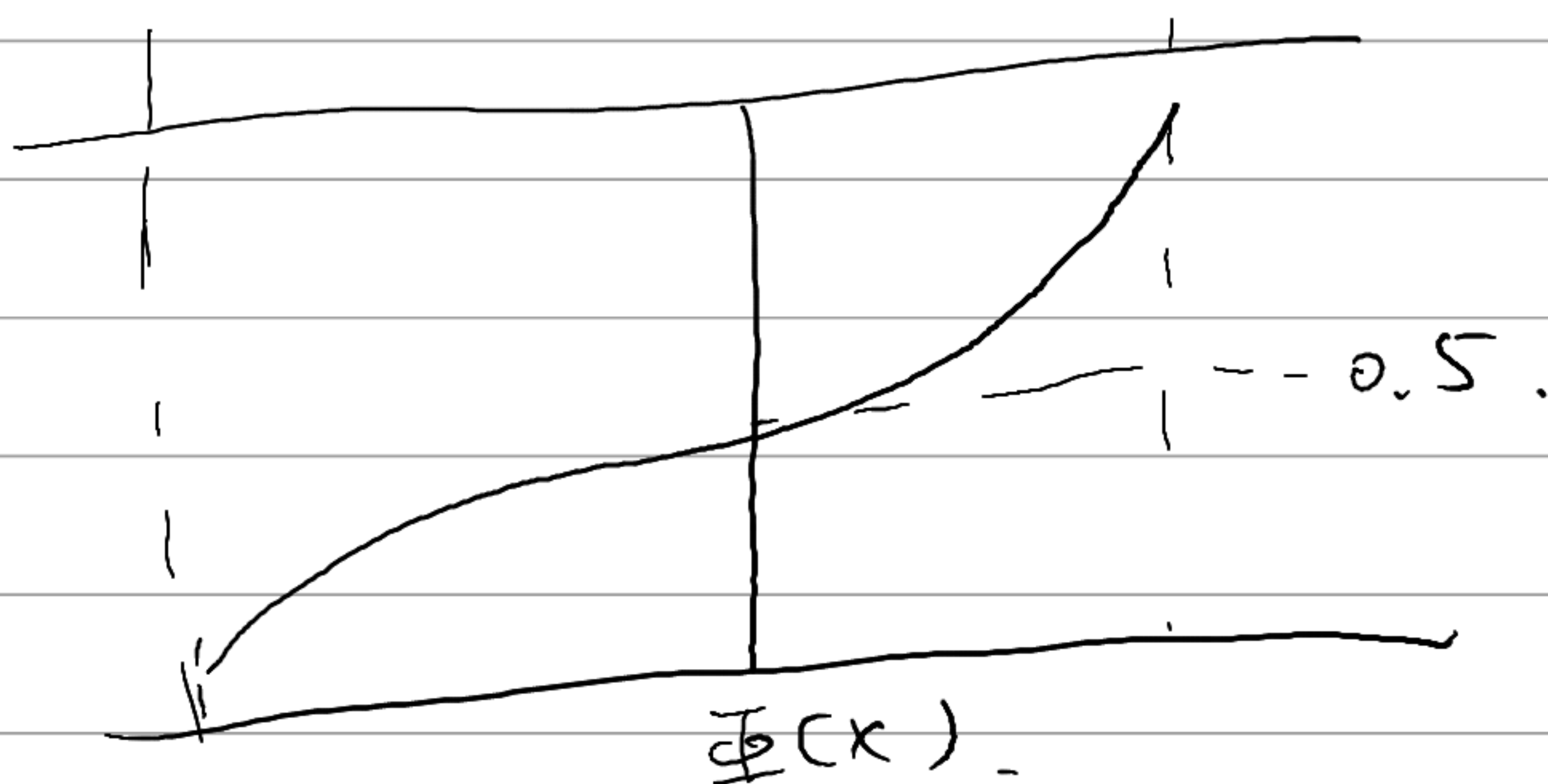
④ + drink + postal + cigarettes.

⑤ zombie + running people.

— Production of GPAs: academic ability, rigor of teaching, time spent, infrastructure, health, social influence, peers.

— Dummy - treatment group, control group.

— Social outcome: frat/sor. Discrete Choice (Probit/Logit)



① Social Influence is present

but we do not know its nature.

② Proximity due to living in a dorm
matters, does not exclude room → room
influence

③ Social effects in social outcomes are more
voluntary.

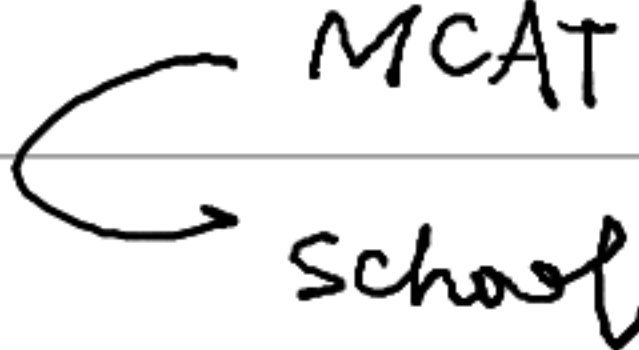
Week 4.

20/02/03

Review: Arcidiacono & Nicholson 2006.

1) Students were chosen, did not choose one another (matching)

2) Positive peer effects

-  MCAT \rightarrow board score
School
(fixed controlled)

\Rightarrow Own effects are present, school peer effects vanish when school fix effects are included.

- Groups of peers: race based, gender based.

Fix \checkmark , MCAT verbal \times female.

- Peer effects on specialty.

Female, high income specialty.

- 1997, 1998, two waves/cohorts

- Keeping up the Joneses.
- Data, American Housing Survey.
 - Dwelling units.
 - Neighborhood : Clusters.
 - 1985, sampled the same cluster in 1989, 1993.
 - Spatial, time info
 - ↳ longitudinal.

- Ioannides (2013), Ch. 2, P14-17

- Canonical Terminology of Manski (1993).

$$y_i = \alpha_0 + \alpha X_i + \overset{\text{Neighborhood's chars.}}{\theta Z_{r(i)}} + \beta E(\text{Neighbors' maintenance})$$

$r(i)$: Neighbor of i .

$$E[y_j | r(i)]$$

$$\Rightarrow y_i = \alpha_0 + \alpha X_i + \theta Z_{r(i)} + \beta E[y_j | r(i)] + \varepsilon_i$$

$E(y_j)$, $j \in r(i)$ are consistent with own decisions, that at equilibrium.

$$\Rightarrow E[y_i] = \alpha_0 + \alpha X_{r(i)} + \theta Z_{r(i)} + \beta E[y_i]$$

$$(1-\beta) E(y_i) = \alpha_0 + \alpha X_{\gamma(i)} + \theta Z_{\gamma(i)}$$

$$E(y_i) = \frac{\alpha_0}{1-\beta} + \frac{\alpha}{1-\beta} X_{\gamma(i)} + \theta \frac{Z_{\gamma(i)}}{1-\beta}$$

Contextual effect

People used to assume that $X_{\gamma(i)} = \text{Neigh/D}$

average of chars. $\equiv Z_{\gamma(i)}$.

$$E(y_i) = \frac{\alpha_0}{1-\beta} + \frac{\alpha + \theta}{1-\beta} Z_{\gamma(i)}. \quad *$$

$$y_i = \alpha_0 + \alpha X_i + \frac{\beta \alpha + \theta}{1-\beta} X_{\gamma(i)} + \varepsilon_i$$

(from rational expectation)

β : coefficient of Neighbors' actions endogenous
social effect.

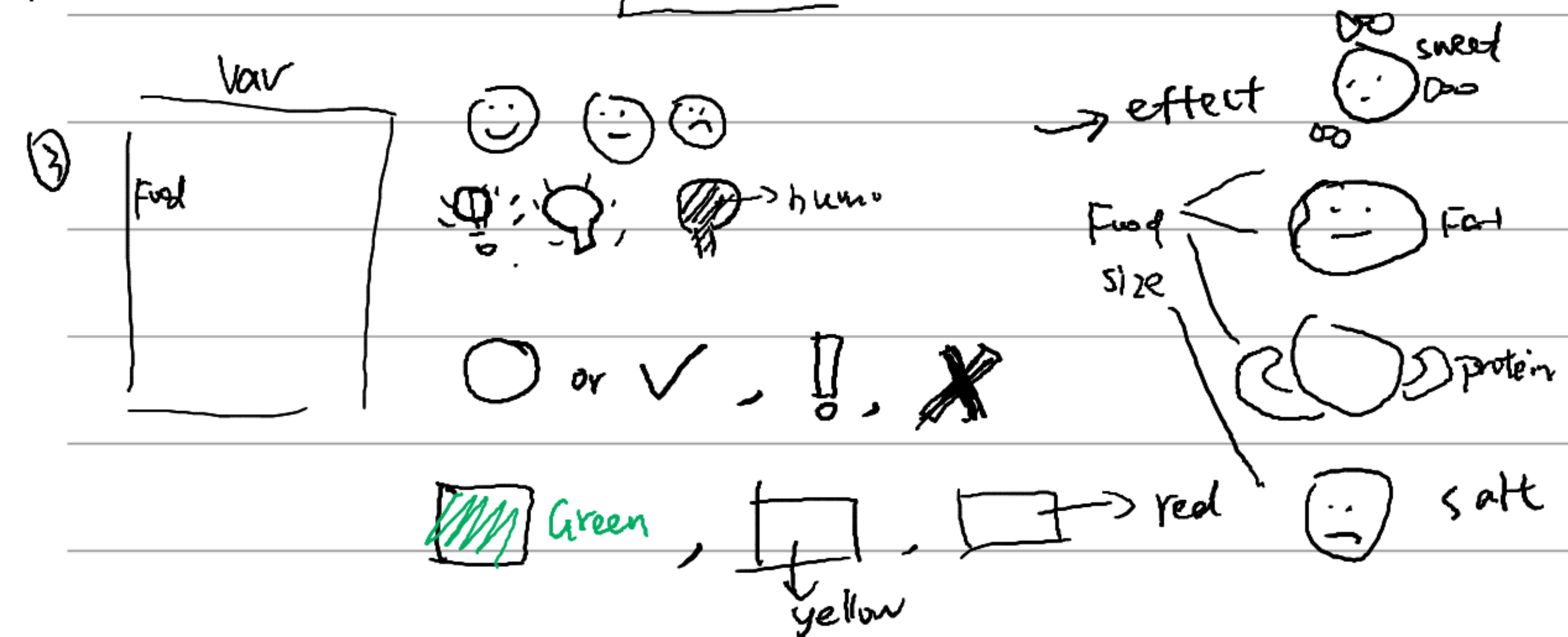
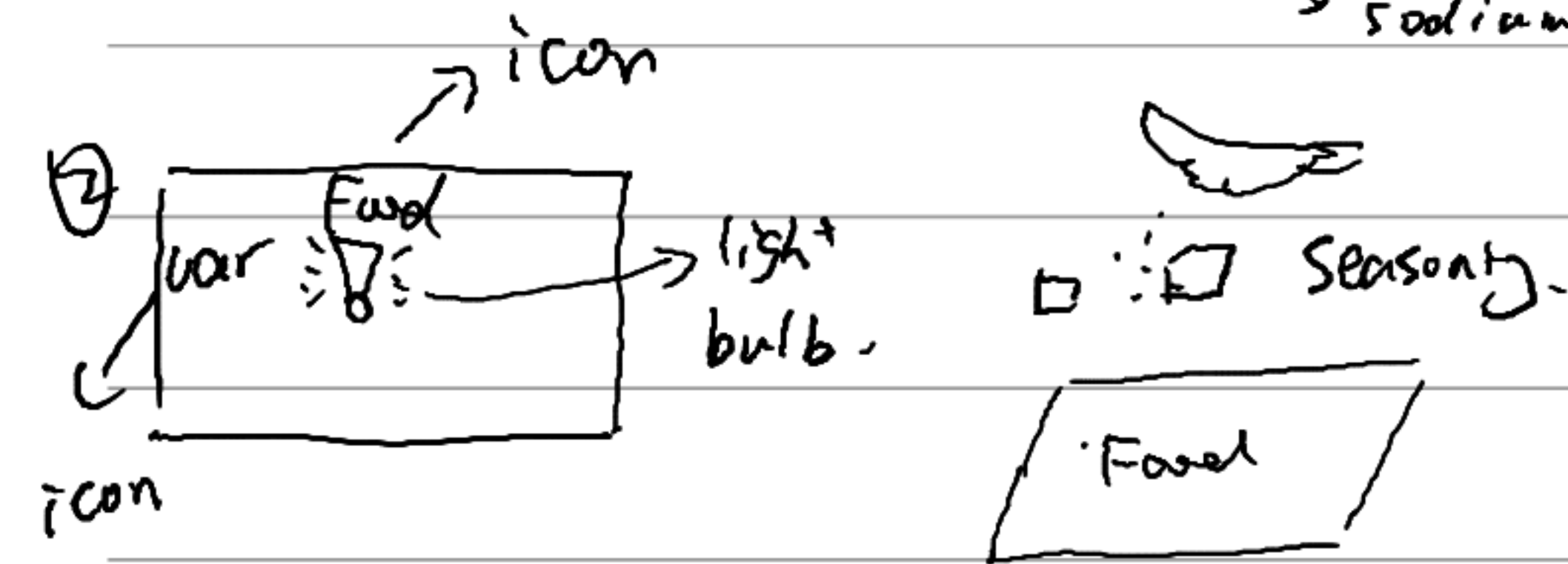
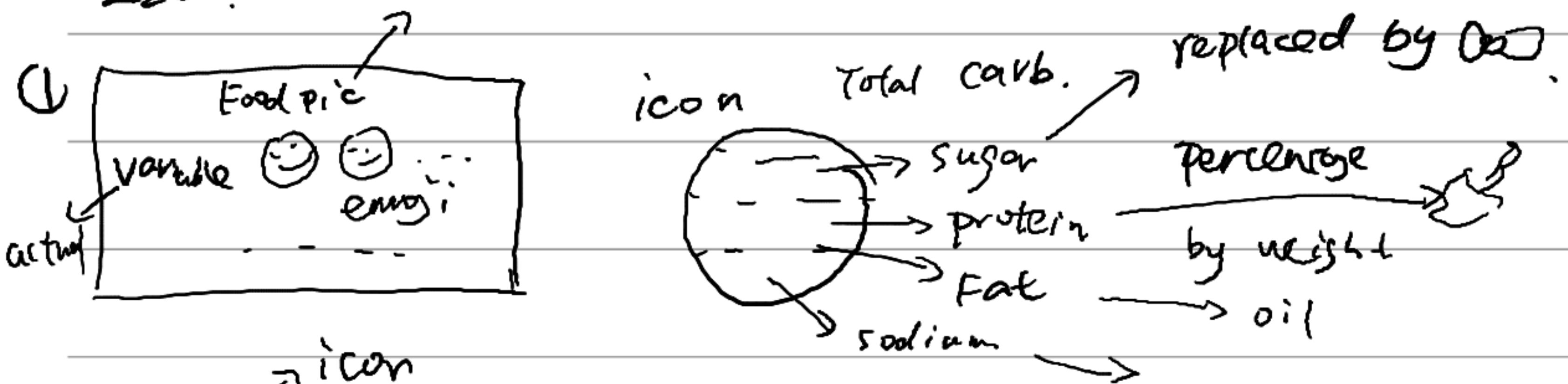
Suppose: $\beta = 0$ (No neighbors' effect).

$$\text{if } \left(\frac{\beta \alpha + \theta}{1-\beta} \right) \neq 0 \Rightarrow \checkmark$$

(oo) Maintenance Expense $\equiv \alpha_0 + \text{location} + \overset{\text{weather}}{\uparrow} \text{degree}$
 $+ \% \text{ ownership} + \# \text{ household white}$
 $+ \text{Pred mean} + \# \text{ vacant} + \text{Quality index} +$
 $\text{Housing's chars.} + \text{occupants' chars.}$

220.

actual / Simple.



A representative food?

① By (size) \leftrightarrow (weight)

② Size \Rightarrow one day percentage.

③

Week 5. L1

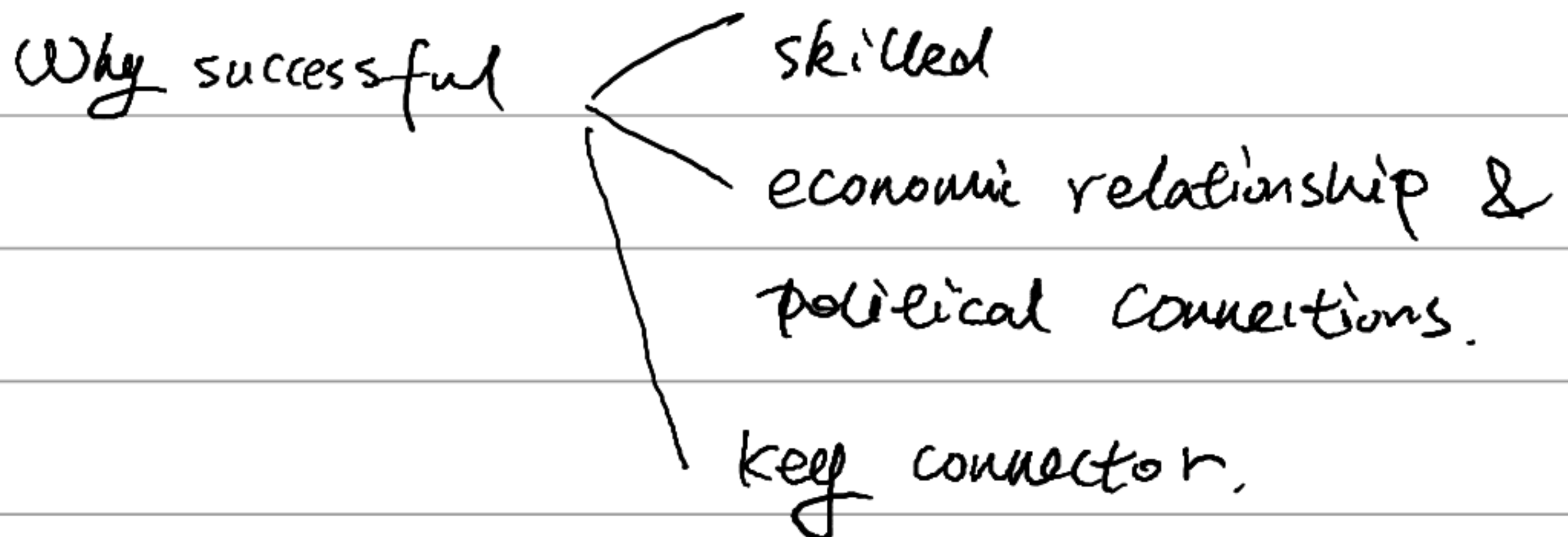
- Back to Networks

- Jackson 2019 P32-40
- Autor et al 2014 P661-666.
- Epidemic Shiller : narrative economics

- Jackson P32-40

• Diffusion. Centrality

Medici Banking & Finance



Power of centrality

• Quantitative measures

- Autor et 2014 RCT

- Randomisations
- Natural Experiments.
- Unanticipated removal of rent control

- Mechanism: direct & indirect effects.

Two types of properties, controlled/never cont.

Rent control: disincentive to landlords to maintain.

Turnover charge.

- Units: houses / condos / cooperative.

Assessment of Housing Values for taxes.

- In MA, local government run school
not county government.

- Model. $\text{value} = f(\text{ ; } A \text{ and } \text{prices})$

1) How well maintained properties are.

2) upper income line nearby.

$$Y_n = C_n + P_n h_n.$$

Cobb-Douglas.

$$C_n = (1 - \alpha) Y_n \quad 0 < \alpha < 1$$

$$U = A C^{1-\alpha} h^\alpha$$

$$P_n h_n = \alpha Y_n$$

Indirect utility function,

$$U_{ID} = A_n (1 - \alpha)^{1-\alpha} Y_n^{1-\alpha} (\alpha Y_n)^\alpha \cdot P_n^{-\alpha}$$

$$= A_n (1 - \alpha)^{1-\alpha} \cdot \alpha^\alpha Y_n P_n^{-\alpha}$$

$$= A_n \cdot \alpha^\alpha Y_n P_n^{-\alpha}$$

$$\bar{U} = A_n \alpha^* y_n p_n^{-\alpha} \rightarrow \text{equilibrium}$$

$$\ln \bar{U} = \ln A_n + \ln \alpha^* + \ln y_n - \alpha \ln p_n.$$

$$\Rightarrow \ln p_n = \frac{1}{\alpha} (\ln A_n + \ln \alpha^* + \ln y_n)$$

Week 6 . 20/02/20 .

• Personnel Economics . → ICPSR UofM

• 1960's Panel Study of Income Dynamics .

- Longitudinal / panel

- Attrition

- Marital, Labor

- Geocodes, Tract / County / State.

↳ 3K-5K people

- Census Tract Data (10 yrs) .

- Moving decisions (what factors affect this)
self-selected

← People with children (chars)
Composition of neighborhoods.

Context

* zoning : Restrictions on land use, Density,
Building ...

* Sorting / Filtering .

* How do people interact with neighbors?

* Mobility, Flexibility

* 同 Survey — self identity → Productivity
female / outcome
under discrimination

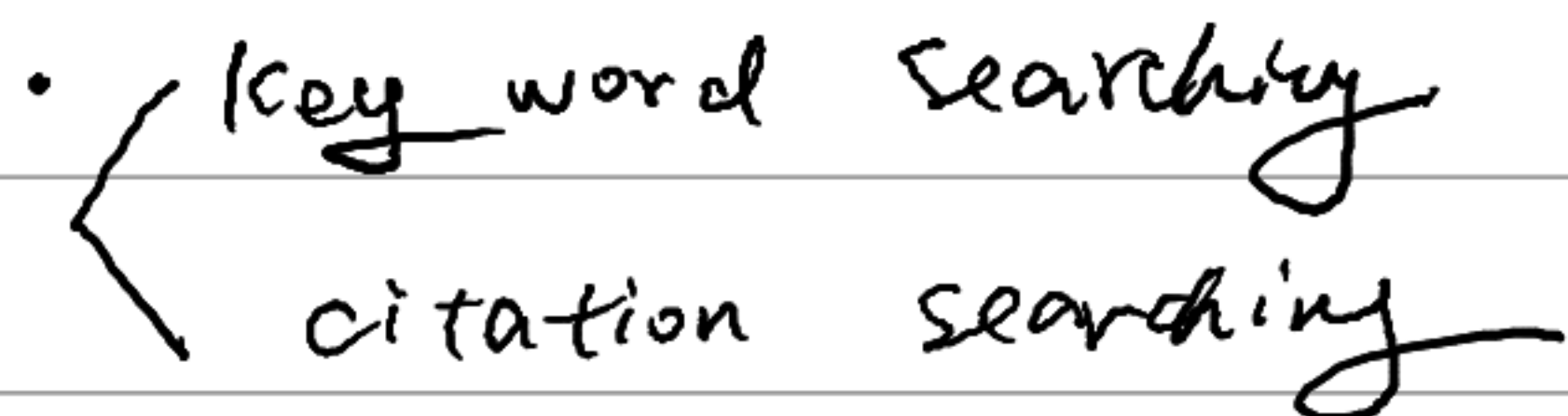
Me! activity

Week 6. 21 Feb 2020 Thursday.

1. Literature Review.

- Annual reviews.

Oxford Bibliographies. (high level explanation)
Dissertation Search.

-  key word searching
citation searching

- Zotero. Citation manager.

2. Datasets.

Census. (social explorer). dataverse

• IPUMS. (microdata)

• ICPSR (researcher data)

3. Policy map (visualization),

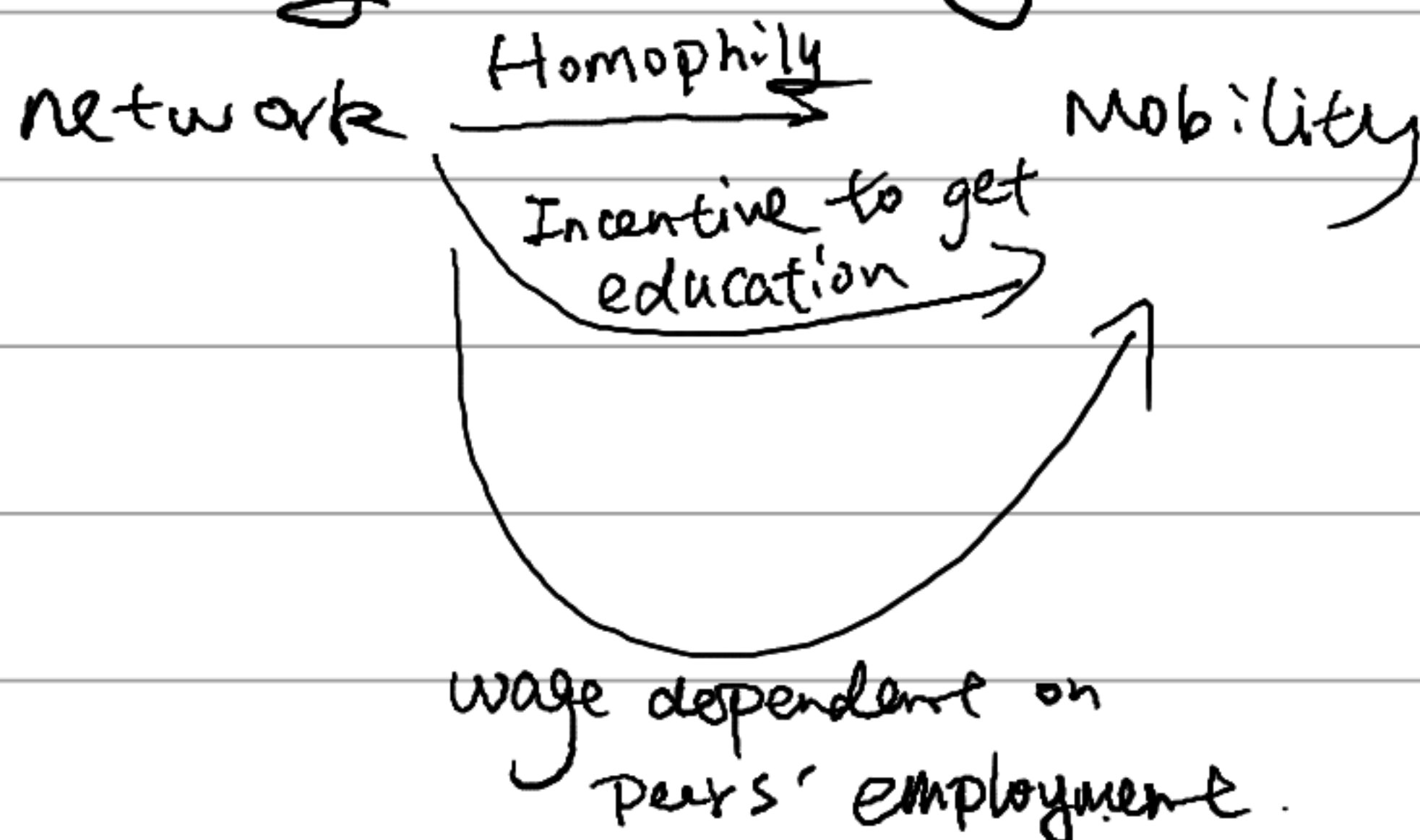
02/26 Wednesday

- Patacchini & Zenou (2011) (Sorting / neighborhood)

- Reading to children (7-31 yrs)
- 2500 children
4900 Parents.
- ↳ Entire cohort.
13 yrs of tracking
- Council House = Public House.
- Questionnaire

- Jackson Book.

- Mobility & Equality



• Intergenerational income elasticity
earning

Raj Chetty on intergenerational mobility

- Gatsby Curve  intertwined.

spatial inequalities.

- Measure inequality.

1) Gini coefficient.

2) US Employment by sector, service sector \leftrightarrow tech

3) Percentage of Children earning more than their parents.

- Education.

1) late 20s, 20% Asian 50% white, 30% black, 20% Hispanic

2) Parents read to babies (# of words, huge gaps)

3) 20% Parents does not know whether they have an impact on the development of children.

4) Social Capital, pass down to parents

Human Capital, resources.

(lower skilled industry, connections to get jobs)

5) Strong Tie / Weak Tie.

Reverse Casualty, Self selection.

6) Moving to opportunities program.

- u : unemployment rate.

v : vacancy rate. (opening).

S : # of contact an individual has.

$(1-u) \cdot v$ Prob of getting info about an opening

$1 - (1-u) \cdot v$ Don't pass info.

Prob that none of S contact pass info

$$(1 - (1-u) \cdot v)^S$$

Prob at least one - - - -

$$1 - (1 - (1-u) \cdot v)^S$$

$S \uparrow$, Prob \uparrow

$v \uparrow$, Prob \uparrow

$u \uparrow$ Prob \downarrow

* S = degree of the network.

20/03/02 Mon.

- Laura Gee 'Social Tie'

☆. Social network → better job match

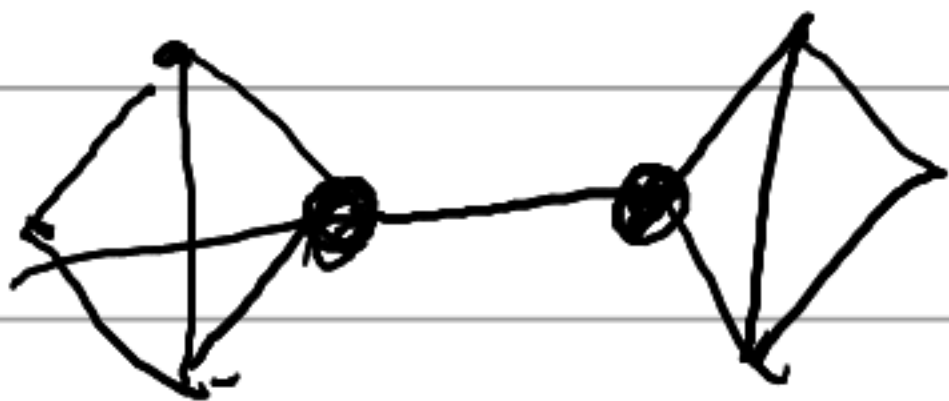
- weak / strong tie. → ↑ productivity
- ↳ shared friends.
- ↳ new info. Subjective report.

Granovetter (1973). Sociologists.

• 90% ties are weak / 90% jobs through a weak tie.
Prob Job ↑ as tie strength ↑

- Novel information
- Identification. (causality)

- Structure based ties / Contact based ties.



- Different Measures of Tie Strength.

Overlaps / Contacts / tag

- Causality, unobservables.

individual (extrovert)
↳ User i's fix effect.
Dyad level

↳ more educated? go to the
same school?

Week 7.

Wednes

03 / March / 2020

- NLSY (US data) Heckman's use of NLSY
- Intermediary.

- Tradeoff \rightarrow intermediaries $\left\{ \begin{array}{l} \text{find job} \\ \text{pay a some percentage} \\ \text{of wage.} \end{array} \right.$

- Cyclicalities in Method of Search.

$$\underline{Y_i} = \alpha_0 + X_i \alpha + \beta \cdot \underbrace{E[Y | x(i)]}_{\downarrow \text{neighborhood effect}} + \underbrace{Z_{x(i)} \theta}_{\downarrow \text{contextual effect}} + \underline{G_i}$$

If $\overline{X}_{x(i)} \equiv Z_{x(i)}$, Reflection Problem

Neighborhood Average of individual characteristics.

• Heckman selection term $E[\varepsilon_i | x_i = \text{Downtown}]$

\rightarrow No longer a reflection problem. $\left\{ \begin{array}{l} \text{self selected.} \end{array} \right.$

Correct self-selection