Pesky irrational people!!!

Classical (Wardrop) user equilibrium (UE) supposes users

- ▶ are all the same (in how they weight the components of virtual cost)
- have perfect information (of the state of the network)
- are perfectly rational

But none of these assumptions are actually true!

New modelling idea: Stochastic User Equilibrium (SUE)

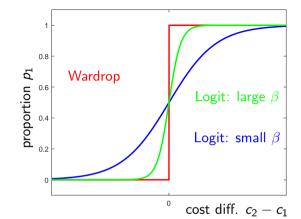
- use randomness to model heterogeneity / imperfect information / irrationality
- ▶ unlike (Wardrop) UE, used choices (i.e., routes) do not have to cost the same

Intro to discrete choice models

- ▶ Simplest setting: binary choice between two alternatives with costs c_1 and c_2
- Basic logit model:

$$p_1=rac{\exp(-eta c_1)}{\exp(-eta c_1)+\exp(-eta c_2)}$$
 and $p_2=rac{\exp(-eta c_2)}{\exp(-eta c_1)+\exp(-eta c_2)}$

- ▶ p_1 , $p_2 > 0$ are the proportions of users taking the two choices. Note $p_1 + p_2 = 1$.
- $\triangleright \beta > 0$ is the sensitivity to cost



- ► Logit also called the **softmax** function
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- ightharpoonup Higher sensitivity eta means a harder max, approaching Wardrop setting
- NB a (very small but nonzero) number of people will make each choice, no matter how bad it is

Discrete choice models: a bit more generally

▶ For *n* choices, so called **multinomial** choice,

$$p_i = \frac{f(c_i)}{\sum_{j=1}^n f(c_j)}$$

will do the 'right' thing, provided f > 0 is a decreasing function of c.

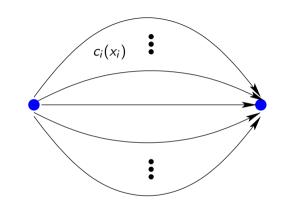
- ▶ For Logit $f(c) = \exp(-\beta c)$.
- ▶ NB automatically each $p_i > 0$ and $\sum_{i=1}^n p_i = 1$
- Partial derivatives:

$$\frac{\partial p_i}{\partial c_i} < 0$$
 and $\frac{\partial p_i}{\partial c_j} > 0 \ (i \neq j)$

(natural interpretation).

Sometimes — choice models expressed in terms of **utility** (usefulness) u_i for each choice — the opposite of cost, e.g. $u_i = -c_i$.

Parallel networks: SUE model



- ▶ Prescribe link cost functions $c_i(x_i)$ for links i = 1, 2, ..., n.
- ▶ Prescribe total demand d.

- ▶ Proportions for each choice are $p_i = x_i/d$.
- ▶ Solve (using Logit with sensitivity $\beta > 0$)

$$\frac{x_i}{d} = \frac{\exp[-\beta c_i(x_i)]}{\sum_{j=1}^n \exp[-\beta c_j(x_j)]}$$

for i = 1, 2, ..., n

- Note $\sum_{i=1}^{n} x_i = d$ automatically from the construction of the system
- ▶ In short *n* (usually, nasty nonlinear) equations in *n* unknowns give it to a computer!

More general networks and SUE

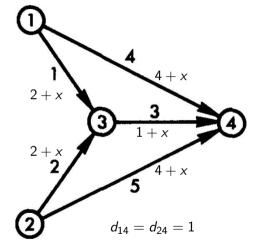
▶ Unlike Wardrop UE problems, there is no 'Beckmann trick', and progress can only be made in the route variables.

► This means:

- You have to compute (all or representatively some of) the routes
- ▶ You have to compute the link-route incidence matrix
- You must express link costs in terms of route variables.
- You then compute route costs in terms of route variables.
- Route costs are input into the choice function.
- ▶ You get large systems of nonlinear equations to solve for the route variables.

▶ In general — massive computation, and no easy examples that can be worked out by hand.

SUE solution — cf complementarity example from before



R1: {4}. R2: {1,3}. R3: {5}. R4: {2,3}

▶ Compute link costs in route variables — formally c(Ay) — then compute route costs in route variables, is $A^{T}c(Ay)$:

R1: $4 + y_1$ R2: $(2 + y_2) + (1 + y_2 + y_4)$

 $= 3 + 2y_2 + y_4$ R3: $4 + y_3$

R4: $3 + y_2 + 2y_4$

▶ (Instead of complementarity) solve:

 $y_1/d_{14} = f(c_{R_1})/[f(c_{R_1}) + f(c_{R_2})]$ $y_2/d_{14} = f(c_{R_2})/[f(c_{R_1}) + f(c_{R_2})]$ $y_3/d_{24} = f(c_{R_3})/[f(c_{R_3}) + f(c_{R_4})]$ $y_4/d_{24} = f(c_{R_4})/[f(c_{R_2}) + f(c_{R_4})]$

▶ Automatically $y_1 + y_2 = d_{14}$ and $y_3 + y_4 = d_{24}$. Unlike Wardop UE, no need to check for unused routes.

Further reading and next steps

- Wikipedia: Discrete choice
- ► Sheffi chapters 10–12
- Warning (1): discrete choice theory should be based formally on random utility theory it gives rise more naturally to Probit choice functions
- Warning (2): random utility theory takes careful account of choices which are not truly independent — e.g., routes with some links in common. This is way too hard to cover in a course like this!!!