

# AAE 637 Lab 9: Conditional logit post-estimation\*

4/15/2015

## Question 1(c)

For each fishing *mode*, you want:

$$P_{ij}\Delta = P_{ij}(\text{cost}_{mode} = \text{cost}_{mode} + 100) - P_{ij}(\text{cost}_{mode} = \text{cost}_{mode}), \forall i, j$$

$$\mathbf{P}\Delta = \begin{bmatrix} P_{11}\Delta & \cdots & P_{14}\Delta \\ \vdots & \ddots & \vdots \\ P_{N1}\Delta & \cdots & P_{N4}\Delta \end{bmatrix}$$

$$P_{ij} = \frac{\exp\{\mathbf{X}_{ij}\boldsymbol{\beta}\}}{\sum_{r=1}^4 \exp\{\mathbf{X}_{ir}\boldsymbol{\beta}\}}$$

$$\mathbf{A} = \begin{bmatrix} \exp\{\mathbf{X}_{11}\boldsymbol{\beta}\} & \cdots & \exp\{\mathbf{X}_{14}\boldsymbol{\beta}\} \\ \vdots & \ddots & \vdots \\ \exp\{\mathbf{X}_{N1}\boldsymbol{\beta}\} & \cdots & \exp\{\mathbf{X}_{N4}\boldsymbol{\beta}\} \end{bmatrix}$$

$\mathbf{A}$  can be constructed by:

```
A = zeros(N,4)
```

```
for j=1:4
```

```
    rhsvar_j = rhsvar(mode_id==j,:);
```

```
    A(:,j) = exp( rhsvar_j * b0 )
```

```
end
```

Performing `sum(A,2)` will yield:

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$$\mathbf{P}_{\text{denom}} = \begin{bmatrix} \sum_{r=1}^4 \exp\{\mathbf{X}_{1r}\boldsymbol{\beta}\} \\ \vdots \\ \sum_{r=1}^4 \exp\{\mathbf{X}_{Nr}\boldsymbol{\beta}\} \end{bmatrix}$$

`P_denom_rep = repmat(P_denom, 1, 4)`

$\mathbf{A} \oslash \mathbf{P}_{\text{denomrep}} = \mathbf{P}$ , where  $\oslash$  is element-by-element division.

To create  $\mathbf{P}\boldsymbol{\Delta}$ , do the same with `costmode = costmode + 100`, perhaps with a different globally-defined `rhsvar` matrix.

Then `mean(P_delta)` will obtain the column means, which is the desired quantity.

For standard errors, since your function above return a vector of 4 elements, you will need to use:

`Grad(params, 'cost_mfx', 4)`

### Question 1(d)

The Cauchy–Schwarz inequality can lead to a shortcut here.

Let  $X$  and  $Y$  be random variables. Then:

$$|E(X \cdot Y)| \leq \sqrt{E(X^2)} \cdot \sqrt{E(Y^2)}$$

This leads to:

$$-\sigma_X\sigma_Y \leq \sigma_{XY} \leq \sigma_X\sigma_Y$$

where  $\sigma_X$  is the standard deviation of  $X$  (defined similarly for  $Y$ ) and  $\sigma_{XY}$  is the covariance of  $X$  and  $Y$

The standard deviation of the difference between two random variables  $X$  and  $Y$  is:

$$\sigma_{X-Y} = \sqrt{\sigma_X^2 + \sigma_Y^2 - 2 \cdot \sigma_{XY}}$$

The null hypothesis of  $X = Y$  is least likely to be rejected when  $\sigma_{X-Y}$  is largest.

Hence, if  $\sigma_{XY}$  is unknown, the “worst case scenario” is where  $\sigma_{XY} = -\sigma_X\sigma_Y$ .

If the null hypothesis can be rejected in the worst case scenario, it can be rejected regardless of the value of the unknown  $\sigma_{XY}$ .

The “worst case scenario” t-stat is:

$$t = \frac{X - Y}{\sqrt{\sigma_X^2 + \sigma_Y^2 - 2 \cdot (-\sigma_X\sigma_Y)}}$$

### Question 1(e)

From Greene 7th ed. page 806, the elasticity of the probability of choosing the  $j$ th alternative with respect to the  $k$ th attribute of choice  $m$  is:

$$\frac{\partial \ln P_{ij}}{\partial \ln x_{mk}} = x_{imk} [\mathbb{1}\{j = m\} - P_{im}] \beta_k$$

$i$  is the observation index.

In our case, catch rate is the 2nd attribute and private boat is the 3rd choice, so  $k = 2$  and  $m = 3$

You want:

$$\begin{bmatrix} \frac{\partial \ln P_{11}}{\partial \ln x_{32}} & \dots & \frac{\partial \ln P_{14}}{\partial \ln x_{32}} \\ \vdots & \ddots & \vdots \\ \frac{\partial \ln P_{N1}}{\partial \ln x_{32}} & \dots & \frac{\partial \ln P_{N4}}{\partial \ln x_{32}} \end{bmatrix}$$

How can we implement this?

Construct  $P_{im}$  in the same way that  $P_{ij}$  was constructed in part (c).

So  $\mathbf{P}$  is  $N \times 4$ .

Then construct  $x_{imk}$  as a matrix:

```
pboat_catch_rate = rhsvar(mode_id==3, 2)
```

```
pboat_catch_rate_mat = repmat(pboat_catch_rate, 1, 4)
```

$\mathbb{1}\{j = m\}$  can be constructed by creating an  $N \times 4$  matrix of zeros and then replacing the 3rd column by ones.

$\beta_k$  is just a scalar, so nothing special is necessary.