

**Estimated Fish Consumption Rates for the U.S.  
Population and Selected Subpopulations  
(NHANES 2003-2010)**

**Appendix C**

**Supplemental Statistical Methodology**

The following provides justification for the assumption that the predicted logit ( $B_{ij}$ ) when including the random effect is proportional to the predicted logit when excluding the random effect ( $B'_{ij}$ ).

Assume that  $C_{ij}$  are realizations from a binomial distribution with probability  $P_{ij}$ :

$$C_{ij} \sim \text{Binomial}(1, P_{ij})$$

The values of  $P_{ij}$  may vary among individuals and be different for the first and second recalls or  $P_{ij}$  may be the same for all individuals and recalls. Regardless of what assumptions are made about  $P_{ij}$ , a basic fact about logistic regression is that when fitting a logistic regression model without random effects or other independent predictors, i.e., fitting the model:

$$\text{Logit}(P_{ij}) = \pi'_0$$

the intercept parameter is equal to:

$$\pi'_0 = \text{Logit}(\text{Mean}(C_{ij})).$$

Since

$$E(\text{Mean}(C_{ij})) = E(C_{ij}) = E(P_{ij}),$$

the intercept is approximately:

$$\pi'_0 \approx \text{Logit}(E(P_{ij})).$$

We can assume a model for the  $P_{ij}$ . For simplicity, assume the probability of consuming fish can be modeled using logistic regression with an intercept, no other predictors, and a random person-specific effect having a normal distribution on the logit scale, i.e.,

$$\text{Logit}(P_{ij}) = \log\left(\frac{P_{ij}}{1 - P_{ij}}\right) = \pi_0 + \pi_i$$

$$\pi_i \sim \text{Normal}(0, \sigma_1^2).$$

We can define the following ratio:

$$\beta = \frac{\pi_0}{\pi'_0}$$

and, with an estimate of  $\beta$  can calculate  $\pi_0$  as:

$$\pi_0 = \beta \pi'_0$$

When the logistic regression model has additional predictors, the predicted logit  $B'_{ij}$  replaces  $\pi'_0$  and the EPA method assumes  $\beta$  is reasonably constant over the range of values of  $B'_{ij}$ .

The following is a heuristic argument for why this assumption is reasonable.

First, let  $P_0 = \text{logistic}(\pi_0)$ . If  $P_{ij}$  is the same for all 24-hour recalls (i.e.,  $P_{ij} = P_0$  and  $\sigma_1^2 = 0$ ) then  $\pi_0 = \pi'_0$  and  $\beta = 1$  for all values of  $\pi_0$ .

If  $\pi_0 = 0$ , the expected probability of fish consumption is  $E(P_{ij}) = 0.50$ , regardless of whether  $\sigma_1^2 = 0$ , i.e.,  $\pi_0 = \pi'_0 = 0$ . If  $\sigma_1^2 = 0$ , all individuals have the same probability of fish consumption and  $E(P_{ij}) = P_0$ . If  $\sigma_1^2 > 0$ , some people have a higher probability and some have a lower probability of fish consumption; however, since the logistic function is symmetric around  $\pi_0 = 0$ , these probabilities balance out and the average probability fish consumption is  $E(P_{ij}) = E(\text{logistic}(\pi_0 + \pi_i)) = E(\text{logistic}(\pi_i)) = 0.50$ . In the case where  $\pi_0 = 0$ ,  $\beta = 0/0$  which is not defined. However, if  $\beta$  is used to define  $\pi_0$  using  $\pi_0 = \beta\pi'_0$ , then any value of  $\beta$  can be used since  $\pi'_0 = 0$ .

Because the logit function is nonlinear and  $P_{ij}$  is limited on the high side (i.e.,  $P_{ij} \leq 1$ ), if  $\pi_0 > 0$ ,  $0.5 < E(P_{ij}) < P_0$ ,  $0 < \pi'_0 < \pi_0$ , and  $\beta > 1$ . Since  $P_{ij}$  is also limited on the low side (i.e.,  $P_{ij} \geq 0$ ), if  $\pi_0 < 0$ ,  $0.5 > E(P_{ij}) > P_0$ ,  $0 > \pi'_0 > \pi_0$ , but because both  $\pi_0$  and  $\pi'_0$  are negative, the ratio is still positive, i.e.,  $\beta > 1$ .  $\beta$  is the same for  $\pi_0$  and  $-\pi_0$ . As  $\pi_0$  increases in absolute magnitude, the non-linearity of the logit function increases. As a result, the difference between  $\pi_0$  and  $\pi'_0$  increases. The EPA method assumes the ratio,  $\beta$ , is relatively constant.

The following provide numerical estimates of  $\beta$ , illustrating the  $\beta$  is reasonably constant for different values of  $\pi_0$  or  $\pi'_0$ .

Given  $\pi_0$  and the variance of the random effect ( $\sigma_1^2$ ), we used numerical integration to calculate  $\beta$ :

$$P_{ij} = \text{logistic}(\pi_0 + \pi_i)$$

$$\beta \approx \frac{\pi_0}{\text{Logit}(E(P_{ij}))}$$

Table C-1 shows  $\beta$  as a function of  $\pi_0$  and  $\sigma_1^2$  calculated using numerical integration.

**Table C-1**  $\beta$  as a function of  $\pi_0$  and  $\sigma_1^2$

	$\pi_0$				
$\sigma_1^2$	-2	-1	0.001	1	2
0.25	1.047	1.056	1.060	1.056	1.047
0.50	1.093	1.107	1.114	1.107	1.093
0.75	1.138	1.156	1.164	1.156	1.138
1.00	1.182	1.202	1.212	1.202	1.182
1.25	1.224	1.246	1.256	1.246	1.224

In Table C-1  $\beta$  is greater than or equal to 1.0, relatively constant across rows corresponding to different values of  $\pi_0$  for the same  $\sigma_1^2$  and increases within increasing  $\sigma_1^2$ . The EPA method does not require that  $\beta$  be constant across all possible values of  $\pi_0$ , but reasonably constant across values of  $B'_{ij}$ .