Recall that

$$p_{jk}^* = P(T^* = t, Z = k)$$
 $p_{jk} = P(T = t, Z = k)$
 $p_k^* = P(T^* = 1 | Z = k)$
 $p_k = P(T = 1 | Z = k)$
 $q = P(Z = 1)$

Thus,

$$p_{00}^* = P(T^* = 0|Z = 0)P(Z = 0) = (1 - p_0^*)(1 - q)$$
$$= \left(\frac{1 - p_0 - \alpha_1}{1 - \alpha_0 - \alpha_1}\right)(1 - q)$$

$$p_{10}^* = P(T^* = 1|Z = 0)P(Z = 0) = p_0^*(1 - q)$$
$$= \left(\frac{p_0 - \alpha_0}{1 - \alpha_0 - \alpha_1}\right)(1 - q)$$

$$p_{01}^* = P(T^* = 0|Z = 1)P(Z = 1) = (1 - p_1^*)q$$
$$= \left(\frac{1 - p_1 - \alpha_1}{1 - \alpha_0 - \alpha_1}\right)q$$

$$p_{11}^* = P(T^* = 1|Z = 1)P(Z = 1) = p_1^*(1 - q)$$
$$= \left(\frac{p_1 - \alpha_0}{1 - \alpha_0 - \alpha_1}\right)q$$

Define

$$F_{tk}^*(\tau) = P(Y \le \tau | T^* = t, Z = k)$$

$$F_{tk}(\tau) = P(Y \le \tau | T = t, Z = k)$$

$$F_k(\tau) = P(Y \le \tau | Z = k)$$

for $t, Z \in \{0, 1\}$. By Bayes' rule, we have

$$F_{0k}(\tau) = \frac{1 - \alpha_0}{1 - p_k} (1 - p_k^*) F_{0k}^*(\tau) + \frac{\alpha_1}{1 - p_k} p_k^* F_{1k}^*(\tau)$$

$$F_{1k}(\tau) = \frac{\alpha_0}{p_k} (1 - p_k^*) F_{0k}^*(\tau) + \frac{1 - \alpha_1}{p_k} p_k^* F_{1k}^*(\tau)$$

Now, the model is $Y = \beta T^* + U$ and

$$F_U(\tau) = P(U \le \tau) = P(Y - \beta T^* \le \tau)$$

but if Z is independent of U then it follows that

$$F_{U}(\tau) = F_{U|Z=k}(\tau) = P(U \le \tau | Z = k) = P(Y \le \tau + \beta T^{*} | Z = k)$$

$$= P(Y \le \tau | T^{*} = 0, Z = k)(1 - p_{k}^{*}) + P(Y \le \tau + \beta | T^{*} = 1, Z = k)p_{k}^{*}$$

$$= (1 - p_{k}^{*})F_{0k}^{*}(\tau) + p_{k}^{*}F_{1k}^{*}(\tau)$$

for all k by the Law of Total Probability. Similarly,

$$F_k(\tau) = (1 - p_k^*) F_{0k}^*(\tau) + p_k^* F_{1k}^*(\tau)$$

and rearranging

$$(1 - p_k^*) F_{0k}^*(\tau) = F_k(\tau) - p_k^* F_{1k}^*(\tau)$$

Substituting this expression into the equation for $F_U(\tau)$ from above, we have

$$F_U(\tau) =$$