

2.) Given the softmax function in (10.13) and (4.13):

(a.) If in (10.13): $f_m(x) = \Pr(Y=m|X) = \frac{e^{z_m}}{\sum_{l=0}^q e^{z_l}}$ \odot

And we add a constant c to each z_l , we get:

$$f_m^*(x) = \frac{e^{z_m+c}}{\sum_{l=0}^q e^{z_l+c}} = \frac{e^{z_m+c}}{\sum_{l=0}^q e^{z_l} e^c} = \frac{e^{z_m} \cdot e^c}{\left(\sum_{l=0}^q e^{z_l}\right) \cdot e^c} = \frac{e^{z_m}}{\sum_{l=0}^q e^{z_l}}$$

which is the same as the above \odot before adding c to each z_l .

(b.) If in (4.13): $\Pr(Y=k|X=x) = \frac{e^{\beta_{k0} + \beta_{k1}x_1 + \dots + \beta_{kp}x_p}}{\sum_{l=1}^K e^{\beta_{l0} + \beta_{l1}x_1 + \dots + \beta_{lp}x_p}} = f$

Then adding constants $c_j, j=0,1,\dots,p$ to the coefficients for each of the classes, then:

$$f^* = \Pr(Y=k|X=x) = \frac{e^{(\beta_{k0}+c_0) + (\beta_{k1}+c_1)x_1 + \dots + (\beta_{kp}+c_p)x_p}}{\sum_{l=1}^K e^{(\beta_{l0}+c_0) + (\beta_{l1}+c_1)x_1 + \dots + (\beta_{lp}+c_p)x_p}}$$

$$= \frac{e^{(\beta_{k0} + \beta_{k1}x_1 + \dots + \beta_{kp}x_p) + (c_0 + c_1x_1 + \dots + c_px_p)}}{\sum_{l=1}^K e^{(\beta_{l0} + \beta_{l1}x_1 + \dots + \beta_{lp}x_p) + (c_0 + c_1x_1 + \dots + c_px_p)}}$$

$$= \frac{e^{c_0 + c_1x_1 + \dots + c_px_p} \cdot e^{\beta_{k0} + \beta_{k1}x_1 + \dots + \beta_{kp}x_p}}{e^{c_0 + c_1x_1 + \dots + c_px_p} \cdot \sum_{l=1}^K e^{\beta_{l0} + \beta_{l1}x_1 + \dots + \beta_{lp}x_p}} = \frac{e^{\beta_{k0} + \beta_{k1}x_1 + \dots + \beta_{kp}x_p}}{\sum_{l=1}^K e^{\beta_{l0} + \beta_{l1}x_1 + \dots + \beta_{lp}x_p}}$$

↓ which is the same as above f .

Thus, when we add constants c_j , predictions at any new point x are unchanged.