### 1 Notation

- n: number of teams;
- N: number of matches;
- $T_k^q$ : last minute of q-th half of match;
- $m_k^q$ : total goals in q-th half of match k;
- $t_{k,l}^q$ : minute of the *l*-th goal in the *q*-th half of match k;
- $t_{k,0}^1=0,\,t_{k,m_k^1+1}^1=T_k^1,\,t_{k,0}^2=45$  and  $t_{k,m_k^2+1}^2=T_k^2;$
- $H_{k,l}^q = \begin{cases} 1, & \text{if the home team scored a goal in interval } [t_{k,l}, t_{k,l+1}); \\ 0, & \text{otherwise}; \end{cases}$
- $A_{k,l}^q = \begin{cases} 1, & \text{if the away team scored a goal in interval } [t_{k,l}, t_{k,l+1}); \\ 0, & \text{otherwise;} \end{cases}$
- $\lambda_k(t)$ : rate of goals for the home team in match k;
- $\mu_k(t)$ : rate of goals for the away team in match k;
- $\rho = [t_k^1, t_k^2, H_k^1, H_k^2, A_k^1, A_k^2];$
- $\Lambda(t_1, t_2) = \int_{t_1}^{t_2} \lambda_k(t) dt$ ;
- $\mathbf{Y}(t_1, t_2) = \int_{t_1}^{t_2} \mu_k(t) \ dt$ .

# 2 Likelihood for match k

In a match k and half q, conditional on the score being (x, y), the distribution of the time to the next home goal is exponential with rate  $\lambda_k(t)$  and the distribution of time to the next away goal is exponential with rate  $\mu_k(t)$ .

So in an interval  $(t_1, t_2]$  of match k, half q and score (x, y), three scenarios can happen:

1. Home team scores one goal

$$P(t, H = 1) = P(\text{home team scores in } (t_1, t_2] \text{ and away team does not})$$
  
=  $\exp{-\mathbf{\Lambda}(t_1, t_2)}\mathbf{\Lambda}(t_1, t_2)\exp{-\mathbf{Y}(t_1, t_2)}$ 

#### 2. Away team scores one goal

$$P(t, A = 1) = P(\text{away team scores in } (t_1, t_2] \text{ and home team does not})$$
  
=  $\exp\{-\mathbf{Y}(t_1, t_2)\}\mathbf{Y}(t_1, t_2)\exp\{-\mathbf{\Lambda}(t_1, t_2)\}$ 

#### 3. Neither team scores

$$P(t, H = 0, A = 0) = \exp{-\Lambda(t_1, t_2)} \exp{-\mathbf{Y}(t_1, t_2)}.$$

We define  $T_k^1=45+U_k^1$  and  $T_k^2=90+U_k^2$  and model  $U_k^1$  and  $U_k^2$  as

$$U_k^1 \sim \text{Poisson}(\pi_1)$$

$$U_k^2 \sim \text{Poisson}(\pi_2)$$

then taking the product over all intervals of both halves of match k, we have

$$\begin{split} L(\rho \mid T_k^1, T_k^2) &= \prod_{l=0}^{m_k^1} \Bigg[ \exp \Bigg\{ - \mathbf{\Lambda}(t_{k,l}^1, t_{k,l+1}^1) \Bigg\} \exp \Bigg\{ - \mathbf{Y}(t_{k,l}^1, t_{k,l+1}^1) \Bigg\} \bigg( \mathbf{\Lambda}(t_{k,l}^1, t_{k,l+1}^1) \bigg)^{H_{k,l}^1} \bigg( \mathbf{Y}(t_{k,l}^1, t_{k,l+1}^1) \bigg)^{A_{k,l}^1} \\ &\prod_{l=0}^{m_k^2} \Bigg[ \exp \Bigg\{ - \mathbf{\Lambda}(t_{k,l}^2, t_{k,l+1}^2) \Bigg\} \exp \Bigg\{ - \mathbf{Y}(t_{k,l}^2, t_{k,l+1}^2) \Bigg\} \bigg( \mathbf{\Lambda}(t_{k,l}^2, t_{k,l+1}^2) \bigg)^{H_{k,l}^2} \bigg( \mathbf{Y}(t_{k,l}^2, t_{k,l+1}^2) \bigg)^{A_{k,l}^2} \Bigg] \end{split}$$

and

$$\begin{split} L(\rho, T_k^1, T_k^2) &= \prod_{l=0}^{m_k^1} \Bigg[ \exp \Bigg\{ - \mathbf{\Lambda}(t_{k,l}^1, t_{k,l+1}^1) \Bigg\} \exp \Bigg\{ - \mathbf{Y}(t_{k,l}^1, t_{k,l+1}^1) \Bigg\} \bigg( \mathbf{\Lambda}(t_{k,l}^1, t_{k,l+1}^1) \bigg)^{H_{k,l}^1} \bigg( \mathbf{Y}(t_{k,l}^1, t_{k,l+1}^1) \bigg)^{A_{k,l}^1} \bigg] \\ &\prod_{l=0}^{m_k^2} \Bigg[ \exp \Bigg\{ - \mathbf{\Lambda}(t_{k,l}^2, t_{k,l+1}^2) \Bigg\} \exp \Bigg\{ - \mathbf{Y}(t_{k,l}^2, t_{k,l+1}^2) \Bigg\} \bigg( \mathbf{\Lambda}(t_{k,l}^2, t_{k,l+1}^2) \bigg)^{H_{k,l}^2} \bigg( \mathbf{Y}(t_{k,l}^2, t_{k,l+1}^2) \bigg)^{A_{k,l}^2} \bigg] \\ &\frac{\pi_1^{U_k^1} \exp \{ - \pi_1 \}}{U_k^1!} \frac{\pi_2^{U_k^2} \exp \{ - \pi_2 \}}{U_k^2!}. \end{split}$$

Finally, the log-likelihood is

$$\begin{split} l(\rho, T_k^1, T_k^2) &= \sum_{l=0}^{m_k^1} \left[ -\mathbf{\Lambda}(t_{k,l}^1, t_{k,l+1}^1) - \mathbf{Y}(t_{k,l}^1, t_{k,l+1}^1) + H_{k,l}^1 \; \log(\mathbf{\Lambda}(t_{k,l}^1, t_{k,l+1}^1)) + A_{k,l}^1 \; \log(\mathbf{Y}(t_{k,l}^1, t_{k,l+1}^1)) \right] \\ &+ \sum_{l=0}^{m_k^2} \left[ -\mathbf{\Lambda}(t_{k,l}^2, t_{k,l+1}^2) - \mathbf{Y}(t_{k,l}^2, t_{k,l+1}^2) + H_{k,l}^2 \; \log(\mathbf{\Lambda}(t_{k,l}^2, t_{k,l+1}^2)) + A_{k,l}^2 \; \log(\mathbf{Y}(t_{k,l}^2, t_{k,l+1}^2)) \right] \\ &+ U_k^1 \log(\pi_1) - \pi_1 - \log(U_k^1!) + U_k^2 \log(\pi_2) - \pi_2 - \log(U_k^2!). \end{split}$$

## 3 DCP

$$\lambda_k(t) = \alpha_i \ \beta_j \ \gamma_h \ \lambda_{xy} \ e^{f(t)\xi_1}$$
$$= \exp\{\alpha_i^* + \beta_i^* + \gamma_h^* + \lambda_{xy}^* + f(t)\xi_1\}$$

where  $\alpha_i^* = \log(\alpha_i)$ ,  $\beta_j^* = \log(\beta_j)$ ,  $\gamma_h^* = \log(\gamma_h)$  and  $\lambda_{xy}^* = \log(\lambda_{xy})$ .

If 
$$f(t) = \log(t+1)$$
 then

$$\Lambda(t_1, t_2) = \int_{t_1}^{t_2} \lambda_k(t) dt = \int_{t_1}^{t_2} \exp\{\alpha_i^* + \beta_j^* + \gamma_h^* + \lambda_{xy}^* + \log(t+1)\xi_1\} dt$$

$$= \exp\{\alpha_i^* + \beta_j^* + \gamma_h^* + \lambda_{xy}^*\} \int_{t_1}^{t_2} \exp\{\log(t+1)\xi_1\} dt$$

$$= \exp\{\alpha_i^* + \beta_j^* + \gamma_h^* + \lambda_{xy}^*\} \int_{t_1}^{t_2} (t+1)^{\xi_1} dt$$

$$= \exp\{\alpha_i^* + \beta_j^* + \gamma_h^* + \lambda_{xy}^*\} \left(\frac{(t_2+1)^{\xi_1+1} - (t_1+1)^{\xi_1+1}}{\xi_1+1}\right)$$

If f(t) = t then

$$\Lambda(t_1, t_2) = \int_{t_1}^{t_2} \lambda_k(t) dt = \int_{t_1}^{t_2} \exp\{\alpha_i^* + \beta_j^* + \gamma_h^* + \lambda_{xy}^* + t\xi_1\} dt$$

$$= \exp\{\alpha_i^* + \beta_j^* + \gamma_h^* + \lambda_{xy}^*\} \int_{t_1}^{t_2} \exp\{t\xi_1\} dt$$

$$= \exp\{\alpha_i^* + \beta_j^* + \gamma_h^* + \lambda_{xy}^*\} \left(\frac{\exp\{t_2\xi_1\} - \exp\{t_1\xi_1\}}{\xi_1}\right)$$

$$= \exp\{\alpha_i^* + \beta_j^* + \gamma_h^* + \lambda_{xy}^*\} \left(\frac{\exp\{t_2\xi_1\} (1 - \exp\{\xi_1(t_1 - t_2)\})}{\xi_1}\right)$$