# Message passing in TrueSkill

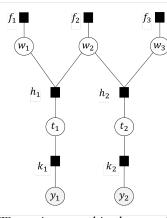
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## Key concepts

- we attempt to apply message passing to TrueSkill
- we encounter two problems
  - the TrueSkill graph isn't a tree
    - we will ignore this problem, but message passing becomes iterative
  - some of the messages don't have standard form
    - approximate using moment matching (seperate chunk)
- we write out messages in excruciating detail

# The full TrueSkill graph



Prior factors:  $f_i(w_i) = \mathcal{N}(w_i; \mu_0, \sigma_0^2)$ 

"Game" factors:

$$h_g(w_{I_g}, w_{J_g}, t_g) = \mathcal{N}(t_g; w_{I_g} - w_{J_g}, 1)$$

$$(I_g \text{ and } J_g \text{ are the players in game } g)$$

Outcome factors:

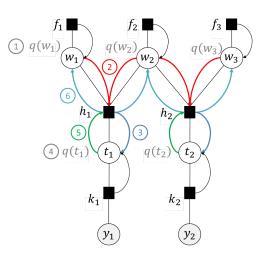
$$k_g(t_g, y_g) = \delta(y_g - sign(t_g))$$



We are interested in the marginal distributions of the skills  $w_i$ .

- What shape do these distributions have?
- We need to make some approximations.
- We will also pretend the structure is a tree (ignore loops).

# Expectation Propagation in the full TrueSkill graph



#### Iterate

- (1) Update skill marginals.
- (2) Compute skill to game messages.
- (3) Compute game to performance messages.
- (4) Approximate performance marginals.
- (5) Compute performance to game messages.
- (6) Compute game to skill messages.

# Message passing for TrueSkill

$$\begin{split} & \prod_{h_g \to w_{I_g}}^{\tau=0}(w_{I_g}) = I, \quad m_{h_g \to w_{J_g}}^{\tau=0}(w_{J_g}) = 1, \quad \forall \ g, \\ & q^{\tau}(w_i) = f(w_i) \prod_{g=1}^{N} m_{h_g \to w_i}^{\tau}(w_i) \sim \mathcal{N}(\mu_i, \sigma_i^2), \\ & m_{w_{I_g} \to h_g}^{\tau}(w_{I_g}) = \frac{q^{\tau}(w_{I_g})}{m_{h_g \to w_{I_g}}^{\tau}(w_{I_g})}, \quad m_{w_{J_g} \to h_g}^{\tau}(w_{J_g}) = \frac{q^{\tau}(w_{J_g})}{m_{h_g \to w_{J_g}}^{\tau}(w_{J_g})}, \\ & = m_{h_g \to t_g}^{\tau}(t_g) = \int h_g(t_g, w_{I_g}, w_{J_g}) m_{w_{I_g} \to h_g}^{\tau}(w_{I_g}) m_{w_{J_g} \to h_g}^{\tau}(w_{J_g}) dw_{I_g} dw_{J_g}, \\ & = q^{\tau+1}(t_g) = \text{Approx} \big( m_{h_g \to t_g}^{\tau}(t_g) m_{k_g \to t_g}(t_g) \big), \quad \text{Need to enforce q to be} \\ & \text{Gaussian so that message} \\ & m_{t_g \to h_g}^{\tau+1}(t_g) = \frac{q^{\tau+1}(t_g)}{m_{h_g \to t_g}^{\tau}(t_g)}, \quad \text{passing can continue to work} \\ & m_{h_g \to w_{I_g}}^{\tau+1}(w_{I_g}) = \iint h_g(t_g, w_{I_g}, w_{J_g}) m_{t_g \to h_g}^{\tau+1}(t_g) m_{w_{I_g} \to h_g}^{\tau}(w_{J_g}) dt_g dw_{J_g}, \\ & m_{h_g \to w_{I_g}}^{\tau+1}(w_{J_g}) = \iint h_g(t_g, w_{J_g}, w_{J_g}) m_{t_g \to h_g}^{\tau+1}(t_g) m_{w_{I_g} \to h_g}^{\tau}(w_{I_g}) dt_g dw_{I_g}. \end{split}$$

### In a little more detail

At iteration  $\tau$  messages m and marginals q are Gaussian, with means  $\mu$ , standard deviations  $\sigma$ , variances  $\nu = \sigma^2$ , precisions  $r = \nu^{-1}$  and natural means  $\lambda = r\mu$ . Step 0 Initialise incoming skill messages:

$$\begin{array}{ccc} r_{\mathbf{h}_g \rightarrow w_i}^{\tau=0} &=& 0 \\ \mu_{\mathbf{h}_g \rightarrow w_i}^{\tau=0} &=& 0 \end{array} \right\} m_{\mathbf{h}_g \rightarrow w_i}^{\tau=0}(w_i)$$

Step 1 Compute marginal skills:

$$\begin{array}{ccc} r_i^{\tau} & = & \\ r_0^{\tau} + \sum_g r_{h_g \to w_i}^{\tau} \\ \lambda_i^{\tau} & = \lambda_0 + \sum_g \lambda_{h_g \to w_i}^{\tau} \end{array} \right\} q^{\tau}(w_i)$$

Step 2 Compute skill to game messages:

$$\begin{array}{ll} r_{w_i \to h_g}^\tau &= r_i^\tau - r_{h_g \to w_i}^\tau \\ \lambda_{w_i \to h_g}^\tau &= \lambda_i^\tau - \lambda_{h_g \to w_i}^\tau \end{array} \right\} m_{w_i \to h_g}^\tau(w_i)$$

#### Step 3 Game to performance messages:

$$\begin{array}{l} \nu_{h_g \rightarrow t_g}^{\tau} \ = \ 1 + \nu_{w_{I_g} \rightarrow h_g}^{\tau} + \nu_{w_{J_g} \rightarrow h_g}^{\tau} \\ \mu_{h_g \rightarrow t_g}^{\tau} \ = \ \mu_{I_g \rightarrow h_g}^{\tau} - \mu_{J_g \rightarrow h_g}^{\tau} \end{array} \right\} m_{h_g \rightarrow t_g}^{\tau}(t_g)$$

#### Step 4 Compute marginal performances:

$$\begin{split} p(t_g) \; &\propto \; \mathcal{N}(\mu_{h_g \to t_g}^{\tau}, \nu_{h_g \to t_g}^{\tau}) \mathbb{I} \big( y - \text{sign}(t) \big) \\ &\simeq \; \mathcal{N}(\tilde{\mu}_g^{\tau+1}, \tilde{\nu}_g^{\tau+1}) \; = \; q^{\tau+1}(t_g) \end{split}$$

We find the parameters of q by moment matching

$$\left. \begin{array}{l} \tilde{\nu}_g^{\tau+1} \; = \; \nu_{h_g \to t_g}^{\tau} \left( 1 - \Lambda \big( \frac{\mu_{h_g \to t_g}^{\tau}}{\sigma_{h_g \to t_g}^{\tau}} \big) \big) \\ \tilde{\mu}_g^{\tau+1} \; = \; \mu_{h_g \to t_g}^{\tau} + \sigma_{h_g \to t_g}^{\tau} \Psi \big( \frac{\mu_{h_g \to t_g}^{\tau}}{\sigma_{h_g \to t_g}^{\tau}} \big) \end{array} \right\} q^{\tau+1}(t_g)$$

where we have defined  $\Psi(x) = \mathcal{N}(x)/\Phi(x)$  and  $\Lambda(x) = \Psi(x)(\Psi(x) + x)$ .

### Step 5 Performance to game message:

$$\left. \begin{array}{ll} \mathbf{r}_{\mathbf{t}_g \rightarrow \mathbf{h}_g}^{\tau+1} &= \tilde{\mathbf{r}}_g^{\tau+1} - \mathbf{r}_{\mathbf{h}_g \rightarrow \mathbf{t}_g}^{\tau} \\ \boldsymbol{\lambda}_{\mathbf{t}_g \rightarrow \mathbf{h}_g}^{\tau+1} &= \tilde{\boldsymbol{\lambda}}_g^{\tau+1} - \boldsymbol{\lambda}_{\mathbf{h}_g \rightarrow \mathbf{t}_g}^{\tau} \end{array} \right\} \mathbf{m}_{\mathbf{t}_g \rightarrow \mathbf{h}_g}^{\tau+1}(\mathbf{t}_g)$$

#### Step 6 Game to skill message:

For player 1 (the winner):

$$\begin{array}{ll} \nu_{h_g \to w_{I_g}}^{\tau+1} &= 1 + \nu_{t_g \to h_g}^{\tau+1} + \nu_{w_{J_g} \to h_g}^{\tau} \\ \mu_{h_g \to w_{I_g}}^{\tau+1} &= \mu_{w_{J_g} \to h_g}^{\tau} + \mu_{t_g \to h_g}^{\tau+1} \end{array} \right\} m_{h_g \to w_{I_g}}^{\tau+1} (w_{I_g})$$

and for player 2 (the looser):

$$\begin{array}{ll} \nu_{h_g \to w_{J_g}}^{\tau+1} &=& 1 + \nu_{t_g \to h_g}^{\tau+1} + \nu_{w_{I_g} \to h_g}^{\tau} \\ \mu_{h_g \to w_{J_g}}^{\tau+1} &=& \mu_{w_{I_g} \to h_g}^{\tau} - \mu_{t_g \to h_g}^{\tau+1} \end{array} \right\} m_{h_g \to w_{J_g}}^{\tau+1} (w_{J_g})$$

Go back to Step 1 with  $\tau := \tau + 1$  (or stop).