

# Message passing in TrueSkill

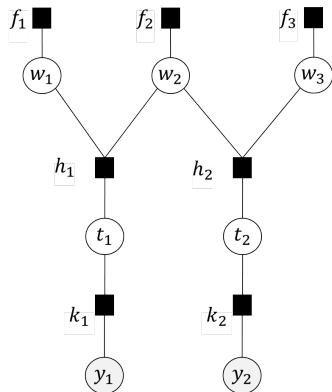
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# Key concepts

- 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:

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

( $I_g$  and  $J_g$  are the players in game  $g$ )

Outcome factors:

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

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).



# Message passing for TrueSkill

$$m_{h_g \rightarrow w_{I_g}}^{\tau=0}(w_{I_g}) = 1, \quad m_{h_g \rightarrow w_{J_g}}^{\tau=0}(w_{J_g}) = 1, \quad \forall g,$$

$$q^\tau(w_i) = f(w_i) \prod_{g=1}^N m_{h_g \rightarrow w_i}^\tau(w_i) \sim \mathcal{N}(\mu_i, \sigma_i^2),$$

$$m_{w_{I_g} \rightarrow h_g}^\tau(w_{I_g}) = \frac{q^\tau(w_{I_g})}{m_{h_g \rightarrow w_{I_g}}^\tau(w_{I_g})}, \quad m_{w_{J_g} \rightarrow h_g}^\tau(w_{J_g}) = \frac{q^\tau(w_{J_g})}{m_{h_g \rightarrow w_{J_g}}^\tau(w_{J_g})},$$

$$m_{h_g \rightarrow t_g}^\tau(t_g) = \iint h_g(t_g, w_{I_g}, w_{J_g}) m_{w_{I_g} \rightarrow h_g}^\tau(w_{I_g}) m_{w_{J_g} \rightarrow h_g}^\tau(w_{J_g}) dw_{I_g} dw_{J_g},$$

$$q^{\tau+1}(t_g) = \text{Approx}(m_{h_g \rightarrow t_g}^\tau(t_g) m_{k_g \rightarrow t_g}(t_g)),$$

$$m_{t_g \rightarrow h_g}^{\tau+1}(t_g) = \frac{q^{\tau+1}(t_g)}{m_{h_g \rightarrow t_g}^\tau(t_g)},$$

$$m_{h_g \rightarrow w_{I_g}}^{\tau+1}(w_{I_g}) = \iint h_g(t_g, w_{I_g}, w_{J_g}) m_{t_g \rightarrow h_g}^{\tau+1}(t_g) m_{w_{J_g} \rightarrow h_g}^\tau(w_{J_g}) dt_g dw_{J_g},$$

$$m_{h_g \rightarrow w_{J_g}}^{\tau+1}(w_{J_g}) = \iint h_g(t_g, w_{I_g}, w_{J_g}) m_{t_g \rightarrow h_g}^{\tau+1}(t_g) m_{w_{I_g} \rightarrow h_g}^\tau(w_{I_g}) dt_g dw_{I_g}.$$

# In a little more detail

At iteration  $\tau$  messages  $m$  and marginals  $q$  are Gaussian, with *means*  $\mu$ , *standard deviations*  $\sigma$ , *variances*  $v = \sigma^2$ , *precisions*  $r = v^{-1}$  and *natural means*  $\lambda = r\mu$ .

Step 0 Initialise incoming skill messages:

$$\left. \begin{aligned} r_{h_g \rightarrow w_i}^{\tau=0} &= 0 \\ \mu_{h_g \rightarrow w_i}^{\tau=0} &= 0 \end{aligned} \right\} m_{h_g \rightarrow w_i}^{\tau=0}(w_i)$$

Step 1 Compute marginal skills:

$$\left. \begin{aligned} r_i^\tau &= r_0 + \sum_g r_{h_g \rightarrow w_i}^\tau \\ \lambda_i^\tau &= \lambda_0 + \sum_g \lambda_{h_g \rightarrow w_i}^\tau \end{aligned} \right\} q^\tau(w_i)$$

Step 2 Compute skill to game messages:

$$\left. \begin{aligned} r_{w_i \rightarrow h_g}^\tau &= r_i^\tau - r_{h_g \rightarrow w_i}^\tau \\ \lambda_{w_i \rightarrow h_g}^\tau &= \lambda_i^\tau - \lambda_{h_g \rightarrow w_i}^\tau \end{aligned} \right\} m_{w_i \rightarrow h_g}^\tau(w_i)$$

### Step 3 Game to performance messages:

$$\left. \begin{aligned} v_{h_g \rightarrow t_g}^\tau &= 1 + v_{w_{I_g} \rightarrow h_g}^\tau + v_{w_{J_g} \rightarrow h_g}^\tau \\ \mu_{h_g \rightarrow t_g}^\tau &= \mu_{I_g \rightarrow h_g}^\tau - \mu_{J_g \rightarrow h_g}^\tau \end{aligned} \right\} m_{h_g \rightarrow t_g}^\tau(t_g)$$

### Step 4 Compute marginal performances:

$$\begin{aligned} p(t_g) &\propto \mathcal{N}(\mu_{h_g \rightarrow t_g}^\tau, v_{h_g \rightarrow t_g}^\tau) \mathbb{I}(y - \text{sign}(t)) \\ &\simeq \mathcal{N}(\tilde{\mu}_g^{\tau+1}, \tilde{v}_g^{\tau+1}) = q^{\tau+1}(t_g) \end{aligned}$$

We find the parameters of  $q$  by *moment matching*

$$\left. \begin{aligned} \tilde{v}_g^{\tau+1} &= v_{h_g \rightarrow t_g}^\tau \left( 1 - \Lambda\left(\frac{\mu_{h_g \rightarrow t_g}^\tau}{\sigma_{h_g \rightarrow t_g}^\tau}\right) \right) \\ \tilde{\mu}_g^{\tau+1} &= \mu_{h_g \rightarrow t_g}^\tau + \sigma_{h_g \rightarrow t_g}^\tau \Psi\left(\frac{\mu_{h_g \rightarrow t_g}^\tau}{\sigma_{h_g \rightarrow t_g}^\tau}\right) \end{aligned} \right\} q^{\tau+1}(t_g)$$

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:

$$\left. \begin{aligned} r_{t_g \rightarrow h_g}^{\tau+1} &= \tilde{r}_g^{\tau+1} - r_{h_g \rightarrow t_g}^{\tau} \\ \lambda_{t_g \rightarrow h_g}^{\tau+1} &= \tilde{\lambda}_g^{\tau+1} - \lambda_{h_g \rightarrow t_g}^{\tau} \end{aligned} \right\} m_{t_g \rightarrow h_g}^{\tau+1}(t_g)$$

### Step 6 Game to skill message:

For player 1 (the winner):

$$\left. \begin{aligned} v_{h_g \rightarrow w_{I_g}}^{\tau+1} &= 1 + v_{t_g \rightarrow h_g}^{\tau+1} + v_{w_{J_g} \rightarrow h_g}^{\tau} \\ \mu_{h_g \rightarrow w_{I_g}}^{\tau+1} &= \mu_{w_{J_g} \rightarrow h_g}^{\tau} + \mu_{t_g \rightarrow h_g}^{\tau+1} \end{aligned} \right\} m_{h_g \rightarrow w_{I_g}}^{\tau+1}(w_{I_g})$$

and for player 2 (the loser):

$$\left. \begin{aligned} v_{h_g \rightarrow w_{J_g}}^{\tau+1} &= 1 + v_{t_g \rightarrow h_g}^{\tau+1} + v_{w_{I_g} \rightarrow h_g}^{\tau} \\ \mu_{h_g \rightarrow w_{J_g}}^{\tau+1} &= \mu_{w_{I_g} \rightarrow h_g}^{\tau} - \mu_{t_g \rightarrow h_g}^{\tau+1} \end{aligned} \right\} m_{h_g \rightarrow w_{J_g}}^{\tau+1}(w_{J_g})$$

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