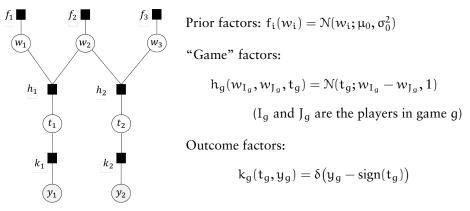
Message passing in TrueSkill

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- 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

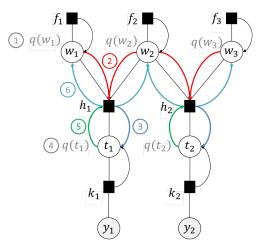
The full TrueSkill graph



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

$$\begin{split} \mathfrak{m}_{h_{g} \to w_{I_{g}}}^{\tau=0}(w_{I_{g}}) &= 1, \quad \mathfrak{m}_{h_{g} \to w_{J_{g}}}^{\tau=0}(w_{J_{g}}) = 1, \quad \forall \ g, \\ q^{\tau}(w_{i}) &= f(w_{i}) \prod_{g=1}^{N} \mathfrak{m}_{h_{g} \to w_{i}}^{\tau}(w_{i}) \sim \mathcal{N}(\mu_{i}, \sigma_{i}^{2}), \\ \mathfrak{m}_{w_{I_{g}} \to h_{g}}^{\tau}(w_{I_{g}}) &= \frac{q^{\tau}(w_{I_{g}})}{\mathfrak{m}_{h_{g} \to w_{I_{g}}}^{\tau}(w_{I_{g}})}, \quad \mathfrak{m}_{w_{J_{g}} \to h_{g}}^{\tau}(w_{J_{g}}) = \frac{q^{\tau}(w_{J_{g}})}{\mathfrak{m}_{h_{g} \to w_{J_{g}}}^{\tau}(w_{I_{g}})}, \\ \mathfrak{m}_{h_{g} \to t_{g}}^{\tau}(t_{g}) &= \iint h_{g}(t_{g}, w_{I_{g}}, w_{J_{g}})\mathfrak{m}_{w_{I_{g}} \to h_{g}}^{\tau}(w_{I_{g}})\mathfrak{m}_{w_{J_{g}} \to h_{g}}^{\tau}(w_{J_{g}})dw_{I_{g}}dw_{J_{g}}, \\ q^{\tau+1}(t_{g}) &= \operatorname{Approx}(\mathfrak{m}_{h_{g} \to t_{g}}^{\tau+1}(t_{g})\mathfrak{m}_{h_{g} \to t_{g}}(t_{g})), \\ \mathfrak{m}_{h_{g} \to w_{I_{g}}}^{\tau+1}(t_{g}) &= \frac{q^{\tau+1}(t_{g})}{\mathfrak{m}_{h_{g} \to t_{g}}^{\tau}(t_{g}}), \\ \mathfrak{m}_{h_{g} \to w_{I_{g}}}^{\tau+1}(w_{I_{g}}) &= \iint h_{g}(t_{g}, w_{I_{g}}, w_{J_{g}})\mathfrak{m}_{t_{g} \to h_{g}}^{\tau+1}(t_{g})\mathfrak{m}_{w_{J_{g}} \to h_{g}}(w_{J_{g}})dt_{g}dw_{J_{g}}, \\ \mathfrak{m}_{h_{g} \to w_{I_{g}}}^{\tau+1}(w_{I_{g}}) &= \iint h_{g}(t_{g}, w_{J_{g}}, w_{J_{g}})\mathfrak{m}_{t_{g} \to h_{g}}^{\tau+1}(t_{g})\mathfrak{m}_{w_{I_{g}} \to h_{g}}(w_{I_{g}})dt_{g}dw_{J_{g}}, \\ \mathfrak{m}_{h_{g} \to w_{J_{g}}}^{\tau+1}(w_{J_{g}}) &= \iint h_{g}(t_{g}, w_{J_{g}}, w_{J_{g}})\mathfrak{m}_{t_{g} \to h_{g}}^{\tau+1}(t_{g})\mathfrak{m}_{w_{I_{g}} \to h_{g}}(w_{I_{g}})dt_{g}dw_{I_{g}}. \end{split}$$

In a little more detail

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

$$\begin{pmatrix} \mathbf{r}_{h_g \to w_i}^{\tau=0} &= 0\\ \mathbf{u}_{h_g \to w_i}^{\tau=0} &= 0 \end{pmatrix} \mathbf{m}_{h_g \to w_i}^{\tau=0}(w_i)$$

Step 1 Compute marginal skills:

$$\left. \begin{array}{l} r_{i}^{\tau} &= r_{0} + \sum_{g} r_{h_{g} \rightarrow w_{i}}^{\tau} \\ \lambda_{i}^{\tau} &= \lambda_{0} + \sum_{g} \lambda_{h_{g} \rightarrow w_{i}}^{\tau} \end{array} \right\} q^{\tau}(w_{i})$$

Step 2 Compute skill to game messages:

$$\begin{array}{l} r^{\tau}_{w_{i} \rightarrow h_{g}} &= r^{\tau}_{i} - r^{\tau}_{h_{g} \rightarrow w_{i}} \\ \lambda^{\tau}_{w_{i} \rightarrow h_{g}} &= \lambda^{\tau}_{i} - \lambda^{\tau}_{h_{g} \rightarrow w_{i}} \end{array} \right\} m^{\tau}_{w_{i} \rightarrow h_{g}}(w_{i})$$

Step 3 Game to performance messages:

$$\left. \begin{array}{l} \nu^{\tau}_{h_g \rightarrow t_g} \ = \ 1 + \nu^{\tau}_{w_{I_g} \rightarrow h_g} + \nu^{\tau}_{w_{J_g} \rightarrow h_g} \\ \mu^{\tau}_{h_g \rightarrow t_g} \ = \ \mu^{\tau}_{I_g \rightarrow h_g} - \mu^{\tau}_{J_g \rightarrow h_g} \end{array} \right\} m^{\tau}_{h_g \rightarrow t_g}(t_g)$$

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

$$\left. \begin{array}{l} \tilde{\nu}_{g}^{\tau+1} \; = \; \nu_{h_{g} \rightarrow t_{g}}^{\tau} \left(1 - \Lambda \left(\frac{\mu_{h_{g} \rightarrow t_{g}}^{\tau}}{\sigma_{h_{g} \rightarrow t_{g}}^{\tau}} \right) \right) \\ \tilde{\mu}_{g}^{\tau+1} \; = \; \mu_{h_{g} \rightarrow t_{g}}^{\tau} + \sigma_{h_{g} \rightarrow t_{g}}^{\tau} \Psi \left(\frac{\mu_{h_{g} \rightarrow t_{g}}^{\tau}}{\sigma_{h_{g} \rightarrow t_{g}}^{\tau}} \right) \end{array} \right\} q^{\tau+1}(t_{g})$$

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:

$$\left. \begin{array}{l} r^{\tau+1}_{t_g \rightarrow h_g} ~=~ \tilde{r}^{\tau+1}_g - r^{\tau}_{h_g \rightarrow t_g} \\ \lambda^{\tau+1}_{t_g \rightarrow h_g} ~=~ \tilde{\lambda}^{\tau+1}_g - \lambda^{\tau}_{h_g \rightarrow t_g} \end{array} \right\} \mathfrak{m}^{\tau+1}_{t_g \rightarrow h_g}(t_g)$$

Step 6 Game to skill message: For player 1 (the winner):

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

and for player 2 (the looser):

$$\begin{pmatrix} \mathbf{v}_{h_g \to \mathbf{w}_{J_g}}^{\tau+1} &= 1 + \mathbf{v}_{t_g \to h_g}^{\tau+1} + \mathbf{v}_{\mathbf{w}_{I_g} \to h_g}^{\tau} \\ \boldsymbol{\mu}_{h_g \to \mathbf{w}_{J_g}}^{\tau+1} &= \boldsymbol{\mu}_{\mathbf{w}_{I_g} \to h_g}^{\tau} - \boldsymbol{\mu}_{t_g \to h_g}^{\tau+1} \end{pmatrix} \mathbf{m}_{h_g \to \mathbf{w}_{J_g}}^{\tau+1}(\mathbf{w}_{J_g})$$

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