

Assessment meets Learning: On the relation between Item Response Theory and Bayesian Knowledge Tracing

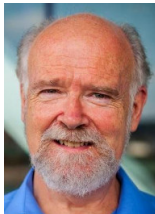
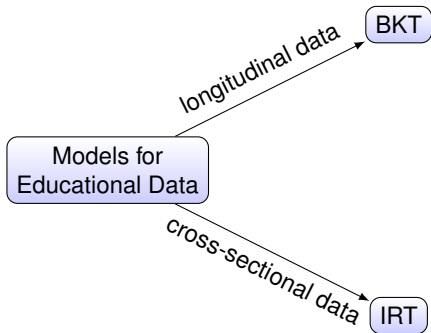
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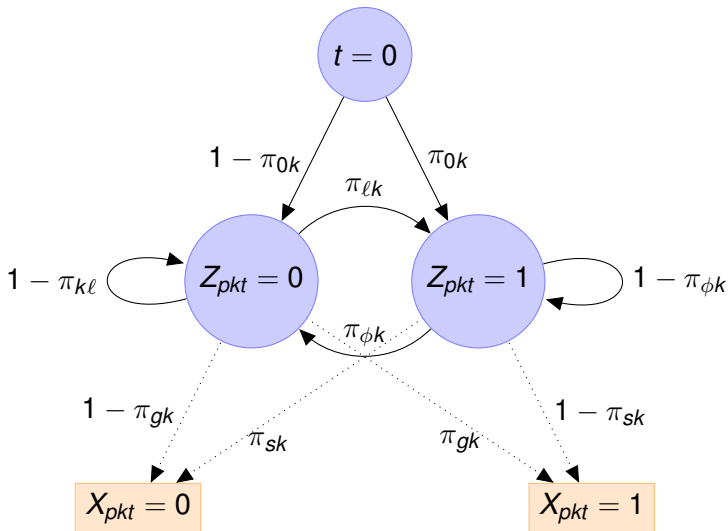
Introduction



Bayesian Knowledge Tracing

- ▶ Standard model used for learning data from intelligent tutoring systems (Corbett and Anderson 1994).
- ▶ Hidden Markov model with binary hidden/latent state and binary emission distribution
- ▶ Standard BKT model utilizes five global (i.e. shared among all individuals) parameters per skill and treats skills independently

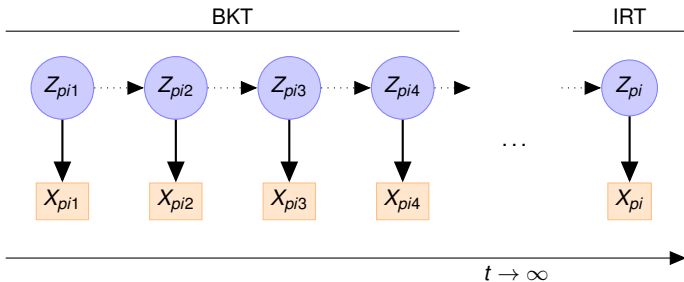
Bayesian Knowledge Tracing



Equilibrium Distribution of Latent Mastery

- ▶ We construct a specific BKT model with person specific learn rate, $\pi_{\ell p}$, item specific forget rate, $\pi_{\phi i}$, and item specific slip and guess parameters π_{si} and π_{gj} .
- ▶ Let X_{pit} be the response of person p to item i at time t .
- ▶ Let Z_{pit} be the latent mastery of person p for item i at time t .

Equilibrium Distribution of Latent Mastery



- ▶ We show that the equilibrium distribution of the latent mastery variable in BKT follows a particular parameterization of an IRT model.
- ▶ Let Z_{pi} be the distribution of Z_{pit} at equilibrium (as $t \rightarrow \infty$). We show

$$\Pr(Z_{pi} = 1) = \frac{\pi_{lp}}{\pi_{lp} + \pi_{\phi i}}$$

Equilibrium Distribution of Latent Mastery

- ▶ Let the transition matrix of the BKT model be denoted by \mathbf{A}_{pi}

$$\mathbf{A}_{pi} = \begin{pmatrix} 1 - \pi_{\phi i} & \pi_{\phi i} \\ \pi_{\ell p} & 1 - \pi_{\ell p} \end{pmatrix}$$

- ▶ Let $\{Z_{pit}\}_{t=1, \dots, T_p}$ denote the Markov chain formed by the transitions of the latent state variable in the BKT model.
- ▶ The stationary distribution for this Markov chain, $\lambda^T = (\lambda_0, \lambda_1)$, exists and is the unique solution which satisfies $\lambda = \mathbf{A}_{pi}^T \lambda$ where $\lambda_0 = P(Z_{pi} = 0)$, $\lambda_1 = P(Z_{pi} = 1)$
- ▶ This stationary distribution can be shown to be

$$\lambda^T = \begin{pmatrix} \frac{\pi_{\phi i}}{\pi_{\ell p} + \pi_{\phi i}} & \frac{\pi_{\ell p}}{\pi_{\ell p} + \pi_{\phi i}} \end{pmatrix}$$

Equilibrium Distribution of Latent Mastery

$$\begin{aligned}P(X_{pi} = 1) &= P(X_{pi} = 1|Z_{pi} = 1)P(Z_{pi} = 1) + \\ &\quad P(X_{pi} = 1|Z_{pi} = 0)P(Z_{pi} = 0) \\ &= (1 - \pi_{si}) \frac{\pi_{lp}}{\pi_{lp} + \pi_{\phi i}} + \pi_{gi} \frac{\pi_{\phi i}}{\pi_{lp} + \pi_{\phi i}}\end{aligned}$$

let $\theta_p = \log \pi_{lp}$ and $\delta_i = \log \pi_{\phi i}$

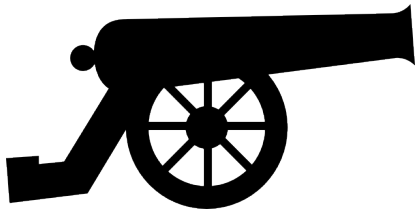
$$\begin{aligned}&= (1 - \pi_{si}) \frac{\exp(\theta_i - \delta_i)}{1 + \exp(\theta_p - \delta_i)} + \pi_{gi} \left(1 - \frac{\exp(\theta_p - \delta_i)}{1 + \exp(\theta_p - \delta_i)} \right) \\ &= \pi_{gi} + ((1 - \pi_{si}) - \pi_{gi}) \frac{\exp(\theta_p - \delta_i)}{1 + \exp(\theta_p - \delta_i)}\end{aligned}$$

The Criticism

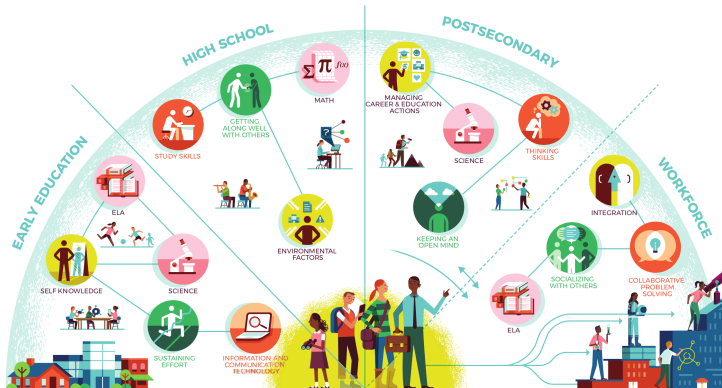
- ▶ No placeholder for education
- ▶ No component in either model to denote teaching or education that is occurring to the learners
- ▶ No component to explain differences in teaching lead to differences in learning outcomes (IRT), or the learning process (BKT).

Ballistic Model

- ▶ BKT model is like firing a canon
- ▶ initial conditions akin to parameters
- ▶ Learning akin to trajectory
- ▶ Reviewers argued role of education in this system is to calibrate such a canon



Holistic Model



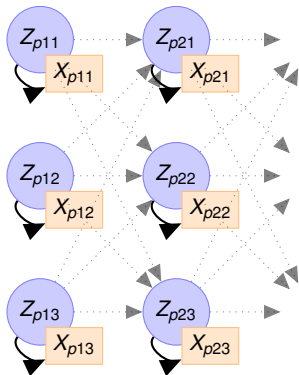
The holistic model of education and work success

Desirable Model Features

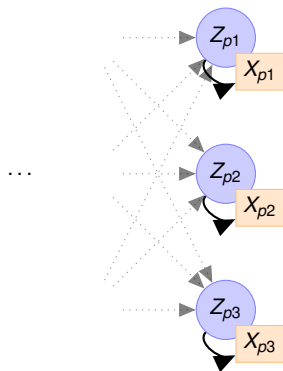
- ▶ Takes hierarchical dependencies into account
- ▶ Take into account well replicated psychological findings
 - ▶ positive manifold
 - ▶ Matthew effect
 - ▶ high predictive validity
- ▶ Should express the dynamic and temporal way that students learn/forget/attain skills/knowledge

Connection to Networks

Interacting Particle System



Ising Network



$t \rightarrow \infty$

Thank You!

- ▶ Collaborators
 - ▶ Michael Yudelson
 - ▶ Maria Bolsinova
 - ▶ Meirav Attali
 - ▶ Gunter Maris
- ▶ Any Questions?

References

- Corbett, A. T. and Anderson, J. R. (1994). "Knowledge tracing: Modeling the acquisition of procedural knowledge". In: *User modeling and user-adapted interaction 4.4*, pp. 253–278.
- Deonovic, B., Yudelson, M., Bolsinova, M., Attali, M., and Maris, G. (2018). "Assessment meets Learning: On the relation between Item Response Theory and Bayesian Knowledge Tracing". In: *ArXiv e-prints*. arXiv: 1803.05926 [stat.ME].