

Generative Models for Ticket Resolution in Expert Networks

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The Life of a Ticket





Application Scenarios





Problem Definition





Outline





Resolution Model (RM)

Routing Algorithms

Each expert has an expertise profile

Generative Models

An expert is likely to be able to resolve tickets similar to what he/she has resolved previously



Tickets resolved by expert E

$$P_{g_i} = [P(w_1|g_i), P(w_2|g_i), ..., P(w_n|g_i)]^T$$

Experiment



Transfer Model (TM)

Routing Algorithms



Generative Models

An expert transfers similar tickets to another expert



Tickets transferred from expert B to expert F

$$P_{e_{ij}} = [P(w_1|e_{ij}), P(w_2|e_{ij}), ..., P(w_n|e_{ij})]^T$$

Experiment

Routing Algorithms

Generative Models



Optimized Network Model (ONM)

Experiment

Transfer profiles optimized for the entire expert network

$$\mathcal{L} = \prod_{t \in T} P(R(t)|t)$$

$$P(R(t)|t) = P(g_1|t)P(g_2|t, g_1)P(g_3|t, g_2)P(g_3|t, g_3)$$

$$P(g_j|t, g_i) = \frac{P(t|e_{ij})P(g_j|g_i)}{Z(t, g_i)}$$

$$= \frac{(\prod_{w_k \in t} P(w_k|e_{ij})^{f(w_k, t)})P(g_j|g_i)}{Z(t, g_i)}$$

$$Z(t, g_i) = \sum_{g_j \in \mathcal{G}} P(t|e_{ij})P(g_j|g_i)$$



Optimized Network Model (ONM)

Generative Models Routing Algorithms Experiment $\log \mathcal{L} \geq \lfloor \log \mathcal{L} \rfloor = \sum \sum (\log(P(t|e_{ij})) + \log(P(g_j|g_i)))$ $e_{ij} t \in T_{ij}$ $-\sum \sum \sum \log(\sum (P(g_{\ell}|g_i) \times P(w_k|e_{i\ell})))$ $g_i \in \mathcal{G} \ t' \in \mathcal{T}_i \ w_k \in t' \qquad g_\ell \in \mathcal{G}$ $\nabla \lfloor \log(\mathcal{L}) \rfloor = \frac{\partial \lfloor \log \mathcal{L} \rfloor}{\partial P(w_k | e_{ij})}$ TM model as initial values Use steepest descent method $= \frac{\sum_{t \in \mathcal{T}_{ij}} n(w_k, t)}{P(w_k | e_{ij})}$ until convergence $P(g_j|g_i) \times \sum_{t' \in T_i} n(w_k, t')$ $\overline{\sum_{g_{\ell} \in \mathcal{G}} P(g_{\ell}|g_i) \times P(w_k|e_{i\ell})}$



Routing Algorithms

- Ranked resolver
- □ Greedy transfer
- Holistic routing



Ranked Resolver



$$P(g_i|t) = \frac{P(g_i)P(t|g_i)}{P(t)} \propto P(g_i) \prod_{w_k \in t} P(w_k|g_i)^{f(w_k,t)}$$



Greedy Transfer

Generative Models

Routing Algorithms

Experiment

Match the ticket with the transfer profiles





Holistic Routing

Generative Models

Routing Algorithms

Experiment

All possibilities are explored





Experimental Results

Generative Models Routing Algorithms Experiment AIX ticket data 18,426 tickets 16,065 words 847 expert groups Evaluation 75% training data 25% testing data

Data items are divided randomly



Experimental Results





Conclusion

- We presented generative models to characterize the ticket resolution process
 - Historical routing sequence and ticket content are integrated together into generative models
 - Both expertise profiles and transfer profiles are captured
 - Model parameters are optimized either locally or globally
- We investigated ticket routing algorithms
 Experiments show that the algorithms are efficient
- Other applications of the generative models
 - Expertise awareness assessment
 - Network organizational structure investigation

Thanks! Questions?