

vertisers who bid for the exact user query can participate in its auctions. However, in real-life systems, a listing can participate in the “related” keywords’ auctions as well. For example, a listing that bids on “shoes” can participate in a user query “running shoes”. We can extend our method to such cases by forming a mapping between the bid keywords and user queries for an advertiser by using logs. In future, we plan to conduct rigorous experiments using our method’s extension for such listings. Finally, we used a first order DLM to capture trends and momentary peaks in query search volume. However, the DLM ignores information about query traffic features and only focuses on the total number of searches. We plan to address this limitation using dynamic Bayes net, that smoothly vary conditional probability distributions over each edge. Bayes net also allows for feature targeting: For example, if the advertiser has a certain geographical area, or a demographic group to which they would like to advertise, then by fixing the corresponding nodes in the Bayes net, we can generate the traffic corresponding to the targeting.

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