

Project October

Read news that you want to read

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1 Abstract

In modern news aggregation services, such as Reddit, Slashdot, Digg, and Hacker News, longtime members report noticing a marked decrease in quality of discourse as the services gain mainstream attention. This gradual but irreversible decline caused by the influx of new users has been dubbed Eternal September. The loss of the longtime members perpetuates the problem, and causes large numbers of excellent contributors to be disenchanted and out-of-place in their own community. We aim to engineer a service that uses technological principles to avoid this, thus improving the user experience and allowing a large community to benefit from thoughtful discourse and interesting articles.

2 October: Hybrid Recommender System

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2.8 **Appendix A**

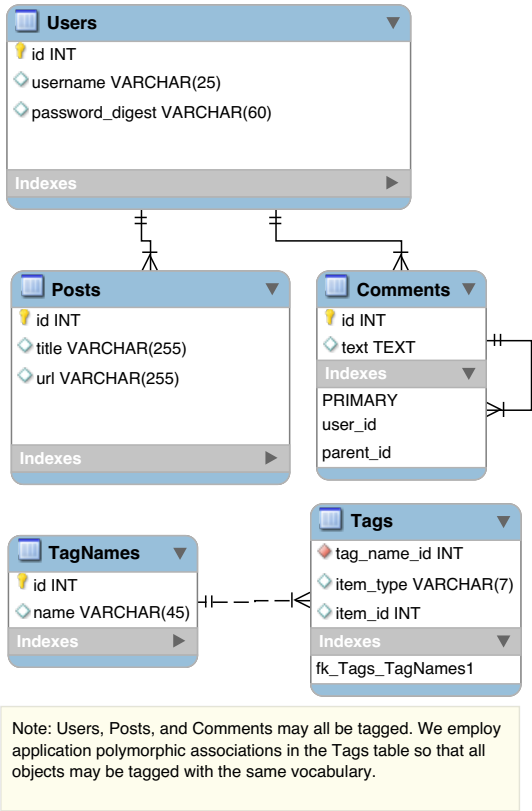


Figure 1: Frontend Database Relational Diagram