

Samantha - the Engine of Data-Driven Intelligent Systems

Introduction

- what's samantha
 - A generic recommender and predictor server. Machine learning is the core, but much more than that, particularly for the purpose of recommendation/ranking/candidate generation/evaluation.
 - MIT licence, oriented to production use (online field experiments in research and typical industrial use)
- the amazing features of samantha, what samantha can do in simple words: full-fledged, self-contained server that can be used in production right away with one configuration file
 - data management, including offline and online, in (indexing) and out (post-processing), bootstrap once for all
 - model management, abstract model operations
 - data processing pipeline, data expanding and feature extraction framework
 - state-of-the-art models: collaborative filtering, matrix factorization, knn, trees, boosting and reinforcement
 - experimental framework for A/B and bucket testing
 - feedback (for online learning/optimization) and evaluation (for experimenting) loop among application front-end, application back-end server and samantha
 - abstract model parameter server, i.e. extensible variable and index spaces
 - generic oracle-based optimization framework/solver with classic solvers
 - flexible model dependency, e.g. model ensemble, stacking, boosting
 - schedulers for regular model rebuilding or backup
 - control and customize all these components through one centralized configuration file
- why do I develop samantha
 - user-centered research
 - offline to online trend
 - lack of good ones, many but no ideal
- the goal (what to support in the near future): the engine of data-driven intelligent systems
 - text modeling
 - model chaining/composition
 - scale up to clusters: learning algorithm and offline data storage
- philosophy of samantha design: “minimum viable product”

Interfaces, Concepts, Architecture and Data Flow

- recommendation/prediction request
- model and config management request
- evaluation request
- data request and subscribers

- schedulers and their runners

Design Patterns/Principles

- dependency injection
 - strongly and weakly follow the pattern
 - inject as needed especially for services
 - actual working class and config class
- singleton
- server-oriented and be mindful of the running phase
 - system initiation
 - request serving

Current Technology Stack

- java, play framework
 - configuration
 - dynamic loading and compiling
 - sbt/scala building
 - handlers vs. system logics
- http server/service-oriented programming
- JSON
- Redis and Elasticsearch
- Database and local file-based storage (extensible to distributed file storage, e.g. HDFS)

Related Work

- compare with recommender tools, lenskit, mahout, libfm etc.
- compare with machine learning tools, weka, libsvm, scikit-learn etc.
- compare with machine learning systems: xgboost, mxnet, etc.
- compare with machine learning server prediction.io

Dive into Samantha

Easy/Minimum Setup (*play in activator*)

Engines and Key Components

- recommender vs. predictor engine
- predictor engine component
 - indexer
 - retriever
 - indexer

- router
- evaluator
- scheduler
- additional recommender component
 - ranker
 - recommender

Indexers for Data In/Out

- JSON, DAO
- real-time data and offline data indexing
- grouped data points output

Retriever for Candidate Generation

Learning Data, Learning Model and Optimization Solver

- oracle-based optimization framework
- parametric model
- loss function: normal pointwise loss and listwise ranking loss
- (online) optimization method using stochastic oracles

Light-weight Feature Extraction Framework

Heavy-weight Data Expanding

Model Service and Management

- model defaults

Flexible Component and Model Dependency

- Component Construction
 - system initiation phase
 - request serving phase
- model ensemble/stacking
- recommender depending on another recommender/reinforcement learning

Evaluator and A/B or Bucket Testing

- Metrics
- Experiment Group Assignment/Tagging

Online Learning and Evaluation Loops

- group tagging
- from data to (feature extracted) instance

Scheduler for Offline Jobs

Extensibility

Distributed Protocol Design and Lightweight Core

Shallow Dependencies/Flattened Software Layers

Extended Interface Protocol through Request

Scalability

Index/Variable Space (Parameter Server)

Extensible DAO

Redis and Elasticsearch Integration

Summary: Production Ready with One Configuration File

Case Studies/Examples

Model and Algorithm Research

- feature-level model ensemble
- model-level model ensemble
- boosting and GB-CENT

Application Integration and User Research

- main integration threads
 - prediction/recommendation request
 - data indexing request
- once-for-all bootstrap
- A/B or bucket testing (between-subjects assignment)

Integrating with Other Systems (Extension/Plugin)

- xgboost
- lenskit
- wikibrain
- etc.

Conclusion and Outlook

- what samantha provides
- what's the roadmap

Appendix

Installation/Setup

List of Models/Algorithms

Dive into Classes/Interfaces

References