Fast Sampling Algorithms for Sparse Latent Variable Models

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Latent Variable Models

- Latent variable models have become a staple of statistical modelling: clustering, topic models, subspace estimation
- Diverse applications range from
 - Organising text documents (e.g. news) and images, to
 - Predicting user behavior (e.g. click patterns), to
 - Targeting ads
- Versatile tools for discovering the hidden thematic structure of objects in a human-understandable format

Global

Share common structure

Global variables: themes that pervade the dataset



Local

Implementation

- Global state typically fits into RAM
- Managing a massive local state
 - Out of core storage
 - Effective since the typical schedule for the Gibbs sampler iterates over the variables in a fixed order



Local variables: labels for data point

Inference Strategy

- Inference is the process of estimating posterior distribution (or the most likely assignment) of all the latent variables
- Unfortunately, the task is intractable and even approximate methods can not be scaled easily due to the data dependencies introduced by global state.



- Leveraging sparsity and better data-structures
 - After gory math, for most Latent variable models the compute step involves drawing from the conditional:

 $p(z_{ij} = k | \text{rest}) \propto \underbrace{n_{ik} f_k(x_{ij})}_{\text{sparse}} + \underbrace{\alpha_k f_k(x_{ij})}_{\text{slowly-varving}}$

- Divide & Conquer! Brute force sampling for sparse term
- Smart tricks for almost constant time sampling like Alias Method or Fenwick tree for global dense slowly varying term
- Lock-free implementation
 - Sampling thread only reads from global, samples and pushes changes to circular buffers
 - Writing thread pulls from the circular buffers and modifies global

Shard global parameters according to x



- MCMC sampling based method
- Our focus because:
- Leverage inherent sparsity present
- Use of smart data-structures for speeding up sampling
- Many works by Blei and Hoffman et.al.
- But updates are dense as it is an expectation over all labels
- Lots of pressure on CPU-RAM bandwidth

Performance

Throughput

Single instance of AWS C4.8xlarge (36 vCPI, 60 GiB RAM)



Accuracy

Per-word log-likelihood after 1000 iteration of each method

Dataset	Tokens	Vanilla LDA	SparseLDA	aliasLDA	F++LDA
ACM	12M	-7.82	-7.81	-7.83	-7.82
NY Times	100M	-7.91	-7.91	-7.92	-7.91
Reuters	105M	-7.34	-7.34	-7.35	-7.35
PubMed	738M	-6.96	-6.92	-6.96	-6.95
Wikipedia	1,418M	-7.58	-7.58	-7.60	-7.59

