The Indian Buffet Process

Eli Bingham¹ Matt Dickenson²

¹University of North Carolina

²Duke University

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Introduction

Indian Buffet Process:

- N customers enter (in sequence) a buffet restaurant with an infinite number of dishes
- $oldsymbol{@}$ First customer fills her plate with Poisson(lpha) number of dishes
- i^{th} customer samples dishes in proportion to their popularity, with probability $\frac{m_k}{i}$, where m_k is the number of previous customers who sampled dish k
- **1** i^{th} customer then samples a Poisson $(\frac{\alpha}{i})$ number of new dishes

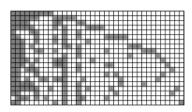


Figure: Griffiths and Ghahramani (2011) Figure 5

Background: Dirichlet Process

Finite version (Dirichlet distribution):

- Assignment of an object to a class is independent of all other assignments: $P(c|\theta) = \prod_{i=1}^{N} P(c_i|\theta) = \prod_{i=1}^{N} \theta_{c_i}$
- $\theta | \alpha \sim \mathsf{Dirichlet}(\frac{\alpha}{K}, \dots, \frac{\alpha}{K})$ (if symettric)
- $c_i | \theta \sim \mathsf{Discrete}(\theta)$, where $\mathsf{Discrete}: \mathsf{Bernoulli}:: \mathsf{Multinomial}: \mathsf{Binomial}$

Integrating out θ : $P(c) = \frac{\prod_{k=1}^{K} \Gamma(m_k + \frac{\alpha}{K})}{\Gamma(\frac{\alpha}{K})^K} \frac{\Gamma(\alpha)}{\Gamma(N+\alpha)}$ What happens as $K \to \infty$? We get the Dirichlet process.



Figure: Griffiths and Ghahramani (2011) Figure 1

Background: Chinese Restaurant Process

- N customers enter (in sequence) a restaurant with an infinite number of tables, each with infinite seating
- ② First customer sits at first table with probability $\frac{\alpha}{\alpha} = 1$
- **3** i^{th} customer sits at the k^{th} table with probability $\frac{m_k}{i+\alpha-1}$, where m_k is the number of previous customers who sat at table k

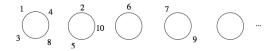


Figure: Griffiths and Ghahramani (2011) Figure 2

Limitation: each object (customer) can only belong to one class (table).

Beta Process

Indian Buffet Process

Demo

Properties of the Resulting Distribution

- The "effective" dimension $K_+ \sim \text{Poisson}(\alpha H_N)$
- The number of dishes on each customer's plate is distributed Poisson(α) (by exchangeability)
- **Z** remains sparse as $K \to \infty$: effective dimensions of **Z** are $N \times K_+$, and the expected number of entries is $N\alpha$

Inference by Gibbs Sampling

Application 1: Choice Behavior

Application 2: Collaborative Filtering

Extension 1: Topic Modeling

Extension 2: Collaborative Filtering

Extensions