

Decompose Restaurant in Detail

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0) Notations

Hierarchical Dirichlet Process Model with Dirichlet-Multinomial:

Hyper-parameter:

α, γ : HDP concentration parameter

λ : Prior for Dirichlet Distribution(W :number of different words, $\phi_1, \dots, \phi_W = \phi$)

Hidden Variable:

(M-step) z : Discrete assignment(t_{ji}, k_{jt} correspond to customer, table assignment in Chinese Restaurant Franchise)

(E-step) θ : Multinomial parameter

Observation:

$x \in (1, \dots, W)$

1) Algorithm: Decompose Restaurant (DR)

GOAL: Maximize log Probability $P = \log p(x, z | \lambda, \alpha, \gamma)$

(i) Make Restaurant j into one table t_0 where customers following uniform distribution:

(% Thus the Probability $P(t_{ji} = t_0) = \frac{1}{W}$)

(ii) Possible Dish = {Nonempty dishes}

(iii) Iterate until no customers are left in this uniform table t_0 :

(a) For each dish $k \in$ Possible Dish, propose to form a new table t_k out of t_0 with dish k and calculate the change ΔP_k :

(% For each customer i in t_0 , sample $t_{ji} \in \{t_0, t_k\} \sim \{\frac{1}{W}, \frac{n_{..k}^w + \phi}{n_{..k} + W\phi}\}$)

(% Propose to form table t_k with customers whose $t_{ji} = t_k$)

(b) Sample a proposal t_{k*} according to the weight and make the new table:

(% Sample a proposal $\{t_{k_1}, \dots, t_{k_K}\} \sim e^{r_{proposal}\{\Delta P_{k_1}, \dots, \Delta P_{k_K}\}}$)

(% $r_{proposal} > 0$, the more decrease of ΔP_k , the less propable to form table t_k)

(c) Possible Dish = Possible Dish / k_*

(iv) TKM: Local search Table/Dish (allowing new dish) + Merge dish

(% Calculate the change of P between present config and config before DR move: ΔP)

(v) Decision:

(% Accept the new config with Probability $\min\{e^{r_{accept}\Delta P}, 1\}$)

(% $r_{accept} > 0$, if P increase, always accept; otherwise more decrease, more likely to reject)