

Bayes net wrapup

- Exact inference algorithms
 - Use to compute $P(X_1, \dots, X_n)$
or $P(X_1, \dots, X_n \mid Y_1, \dots, Y_m)$
- Approximate inference algorithms
 - Direct sampling
 - Rejection sampling
 - Likelihood weighting

Direct Sampling

- To estimate $P(X_1, \dots, X_n)$, sample from the bayes net and count how many samples match the query.
- Divide by the total number of samples.

Rejection Sampling

- We want to estimate $P(X_1, \dots, X_n \mid Y_1, \dots, Y_m)$.
- $P(X_1, \dots, X_n \mid Y_1, \dots, Y_m)$
 $= P(X_1, \dots, X_n, Y_1, \dots, Y_m) / P(Y_1, \dots, Y_m)$
- To estimate this conditional probability, sample from the bayes net and count how many samples match the numerator, divide by the number of samples that match the denominator.
- Called rejection sampling because if the Y variables are very restrictive, not many samples will match (we reject them).

Likelihood weighting

- We want to estimate $P(X_1, \dots, X_n \mid Y_1, \dots, Y_m)$.
- We avoid generating samples that don't fit the Y variables.
 - Rejection sampling wastes time by throwing these out.
- Instead, we directly generate samples that fit the Y variables along with a corresponding **weight**.
- After generating a bunch of samples, count them proportionally to the sum of their weights.