

Probability/Preparation

- basic notions
 - {not, and, or}; conditional probability; independence
 - random variables: X , $f(x)$, $F(x)$, $\mathbb{E}(X)$, $\text{Var}(X)$
- harder ideas (takes time and experience to master)
 - $f_{XY}(x, y) \rightarrow f_X(x) \rightarrow f_{Y|X}(y|x)$; covariance
 - law of total $\{\mathbb{P}(\dots), \mathbb{E}(\dots), \text{Var}(\dots)\}$
- two specific problems (important for statistics)
 - distributions of $g(X, Y)$, $h(X_1, X_2, \dots, X_n)$
 - compound distributions (hierarchically specified ...
... models)

Frequentist Estimation

- estimator
 - a random variable in itself
 - referring to the **procedure** of estimation
 - rather than the **numeric value**
- finding estimators
 - method of moments; maximum likelihood
- assessing estimators
 - sampling distribution
 - bias, variance, mean squared error

Bayesian Estimation

- basic set-up:
 - $\pi(\theta|\text{data}) = f(\text{data}|\theta)\pi(\theta)/m(\text{data})$
- simultaneous estimation problem: (h for “hyperparameter”)
 - $X_i \sim f(\cdot|\theta_i)$ independently for $i = 1, \dots, n$, and $\theta_i \stackrel{iid}{\sim} \pi(\cdot|h)$
- simple Bayes: (h non-random, specified)
 - specify h , and use $\pi(\theta_i|x_i, h) = f(x_i|\theta_i)\pi(\theta_i|h)/m(x_i|h)$
- empirical Bayes: (h non-random, estimated)
 - estimate h w/ all $X_i \stackrel{iid}{\sim} m(\cdot|h)$, and plug into $\pi(\theta_i|x_i, h)$
- hierarchical Bayes: (h random)
 - put prior on h , and go for $\pi(\theta_1, \dots, \theta_n, h|x_1, \dots, x_n)$

Confidence Sets

- pivotal quantity $h(\theta; \mathbf{D})$
 - must know its distribution (not easy)
 - e.g., $T(\mu; \mathbf{D}) \equiv (\bar{X} - \mu) / \sqrt{S^2/n}$
- e.g., likelihood ratio $\Lambda(\theta; \mathbf{D})$
 - $2 \log \Lambda(\theta; \mathbf{D})$, and the χ^2 -distribution

Significance Tests

- key jargon
 - null hypothesis, test statistic, significance level, p-value
- test statistic vs. pivotal quantity
 - $\mathbb{P}[h(\theta; \mathbf{D}) \in C_\alpha] = 1 - \alpha \Rightarrow \mathbb{P}[h(\theta_0; \mathbf{D}) \notin C_\alpha] \stackrel{H_0}{=} \alpha$
 $\underbrace{\hspace{10em}}_{\substack{\{\theta: h(\theta; \mathbf{D}) \in C_\alpha\} \text{ is} \\ \text{a confidence set}}} \quad \underbrace{\hspace{10em}}_{\substack{\theta_0 \text{ rejected} \\ \notin \text{confidence set}}}$
- some specific tests
 - likelihood-ratio test (LRT), t -test, (later) F -test
- multiple testing problems
 - Bonferroni, Benjamini-Hockberg

Computational Techniques

- EM algorithm (basically, a “fancy” MLE algorithm)
 - useful for problems with missing/latent data
- Gibbs sampler
 - sequentially draw from conditional distributions
 - useful for intractable posterior distributions
- Bootstrap
 - $h(\hat{\theta}; \mathbf{D}^{*(1)}), h(\hat{\theta}; \mathbf{D}^{*(2)}), \dots, h(\hat{\theta}; \mathbf{D}^{*(B)})$
 - approximates the distribution of $h(\theta; \mathbf{D})$
 - versatile, can replicate “anything” we do on \mathbf{D}

Linear Models

- in terms of **ideas**, nothing new
 - e.g., **MLE**, **bias**, **variance**, **testing** (t , F)
- in terms of **techniques** (i.e., math), mostly linear algebra
 - $\text{span}(\mathbf{X}) \subset \mathbb{R}^n$, projection, hat matrix \mathbf{H}
- model assessment and choice
 - F -test (nested models), CV, GCV, AIC
- ridge regression
 - variance reduction, multicollinearity
- generalized linear models (GLMs)
 - exponential families, Newton-Raphson (IRLS)

Experimental Design

- some important **ideas** + jargon
 - observational vs. experimental studies
 - association vs. causation
 - confounding, randomization, blocking
- in terms of **techniques**, nothing new (again)
 - ANOVA; *F*-tests — just nested linear models

Survey Sampling

- in terms of **ideas**, nothing new (again)
 - e.g., **bias**, **variance**
- in terms of **techniques**, just one new-ish “trick”
 - $I(x_j \in \mathcal{S}), I(x_j, x_\ell \in \mathcal{S}); \mathbb{P}(x_j \in \mathcal{S}), \mathbb{P}(x_j, x_\ell \in \mathcal{S})$
 - Horvitz-Thompson, inverse probability weighting
- in terms of methods, three basic strategies
 - simple random sampling
 - stratified sampling
 - cluster sampling (brief mention only)

Final Examination

(December 8, 2023; 16:00 – 18:30)

- Q1 (25 pts): matching and multiple choices on various topics
- Q2 (15 pts): Bayesian inference
- Q3 (18 pts): EM algorithm
- Q4 (20 pts): linear model + significance tests
- Q5 (12 pts): confidence interval
- Q6 (10 pts): survey sampling

Remark

Questions are NOT ordered by difficulty—e.g., for Q1 the questions are arranged by spacing and layout considerations alone.

Office Hours Next Week

Monday, **December 4**, 2023

12:00 – 13:00; M3 2103B
(usual)

14:00 – 18:00; M3 4007[†]
(extra)

Wednesday, **December 6**, 2023

10:00 – 16:00; M3 4007[†]
(extra)

[†]with some coffee breaks in between

Feedback

- official UW course feedback
 - completely anonymous
 - <https://perceptions.uwaterloo.ca>
 - by 23:59 EST, Tuesday, **December 5**, 2023
- *Essential Statistics* book
 - please consider writing a review
 - comments, errors, typos, ... please email