Sociology 740 John Fox

Lecture Notes

Review for the Second Exam

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Second Review

1. Diagnostics: Unusual and Influential Data

- ► Outliers, leverage and influence.
- ► Assessing leverage: Hat-values.
- ▶ Detecting outliers: Studentized residuals.
- ▶ Measuring influence: Influence on coefficients and Cook's *D*.
- ▶ Joint influence: Added-variable plots.

2. Diagnostics: Collinearity and Model Selection

- ▶ Nature of the problem.
- ▶ Variance-inflation factors (VIF) and generalized variance-inflation factors (GVIF).
- ▶ Putative solutions:
 - model respecification.
 - variable selection.
 - biased estimation.
 - prior information.
- ▶ Model selection criteria: Mallows's C_p , cross-validation, generalized cross-validation, AIC, BIC.
- Model validation.

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3. Diagnostics: Non-Normality, Non-Constant Error Variance, and Nonlinearity

- ► Non-normality: Quantile-comparison plot, histogram or density estimate; boxplot; transformations.
- ▶ Non-constant error variance: Plotting residuals against fitted values; spread-level plot; transformations, WLS regression, and "corrected" standard errors.
- ▶ Nonlinearity: Component+residual plots; transformations, polynomial regression, and regresson splines.
- ▶ Discrete data: testing for nonlinearity ("lack of fit") and non-constant error variance (Levene's test).

- ► Maximum-likelihood methods (treat as optional):
 - Box-Cox transformation of Y.
 - Box-Tidwell transformation of the X's.
 - constructed variables and score tests.
 - score test for non-constant error variance.

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4. Logit and Probit Models for Dichotomous Data

- ▶ Linear probability, logit, and probit models for dichotomous data.
- ▶ Interpretation of coefficients in the logit model:
 - $B_j/4$ is the effect on the estimated probability of "success" $\widehat{\pi}$ of increasing X_j by 1 (or, for a dummy variable, in comparison to the baseline category), holding other Xs constant, when $\widehat{\pi}$ remains near .5.
 - $\exp(B_j) = e^{B_j}$ is the *multiplicative* effect on the estimated odds of "success" $\widehat{\pi}/(1-\widehat{\pi})$ of increasing X_j by 1 holding other Xs constant.
- ▶ Wald and likelihood-ratio tests; analysis of deviance.

5. Logit and Probit Models for Polytomous Data

- ▶ Polytomous (multinomial) logit model.
- ▶ Nested dichotomies.
- ▶ Proportional-odds model (ordered logit model).

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6. Generalized Linear Models

- ► Format of GLMs:
 - conditional distribution of *Y*:
 - exponential families: Gaussian, binomial, Poisson, gamma, inverse-Gaussian — fit by ML.
 - others: quasi-binomial, quasi-Poisson (for overdispersed binomial and Poisson data) — fit by quasi-likelihood.
 - dispersion parameter ϕ and conditional variance of Y.

| Family | Canonical Link (see below) | Range of Y_i | $V(Y_i \eta_i)$ |
|------------------|----------------------------|------------------------|---|
| Gaussian | identity | $(-\infty, +\infty)$ | ϕ |
| binomial | logit | $\frac{0,1,,n_i}{n_i}$ | $\mu_i(1-\mu_i)$ |
| Poisson | log | $0, 1, 2, \dots$ | μ_i |
| gamma | inverse | $(0,\infty)$ | $egin{array}{c} \mu_i \ \phi \mu_i^2 \end{array}$ |
| inverse gaussian | inverse-square | $(0,\infty)$ | $\phi\mu_i^3$ |

- linear predictor: $\eta_i = \alpha + \beta_1 X_{i1} + \cdots + \beta_k X_{ik}$.
- ullet link function, $g(\mu_i)=\eta_i$; and inverse-link (mean) function, $g^{-1}(\eta_i)=\mu_i$.

| Link | $\eta_i = g(\mu_i)$ | $\mu_i = g^{-1}(\eta_i)$ |
|-----------------------|--------------------------------|---------------------------|
| identity | μ_i | η_i |
| log | $\log_e \mu_i$ | e^{η_i} |
| inverse | $\log_e \mu_i \\ \mu_i^{-1}$ | η_i^{-1} |
| inverse-square | μ_i^{-2} | $\eta_i^{-1/2}$ |
| square-root | $\sqrt{\mu_i}$ | η_i^2 |
| logit | $\log_e \frac{\mu_i}{1-\mu_i}$ | $\frac{1}{1+e^{-\eta_i}}$ |
| probit | $\Phi^{-1}(\mu_i)$ | $\Phi(\eta_i)$ |
| log-log | $-\log_e[-\log_e(\mu_i)]$ | $\exp[-\exp(-\eta_i)]$ |
| complementary log-log | $\log_e[-\log_e(1-\mu_i)]$ | $1 - \exp[-\exp(\eta_i)]$ |

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- ▶ Poisson and quasi-Poisson models for count data
 - Interpretation of coefficients: $e(B_j) = \exp(B_j)$ is the multiplicative effect on expected response count of increasing X_j by 1 (or, for dummy variable, in comparison to baseline category), holding other Xs constant.
 - Same estimated coefficients for Poisson and quasi-Poisson models, but SEs for quasi-Poisson model multiply by $\sqrt{\widehat{\phi}}$ (and thus are typically larger).
- ► Analysis of deviance.
- ▶ Diagnostics: studentized residuals, hat-values, Cook's D, dfbeta and dfbetas, added-variable plots, component+residual plots.

7. Overview of Linear and Generalized-Linear Models

| Explanatory | Response | Type of Model | |
|--------------------------|-------------------------|-------------------------|--|
| Variables | Variable | | |
| Quantitative | Quantitative | Regression | |
| (e.g., education, years) | (e.g., income, dollars) | Regression | |
| Categorical | Quantitative | Analysis of Variance | |
| (e.g., region, gender) | Quantitative | Alialysis of Vallatice | |
| | | Dummy Regression/ | |
| Mixed | Quantitative | Analysis of Covariance/ | |
| | | General Linear Model | |

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| Explanatory Variables | Response Variable | Type of Model |
|--------------------------|---|--|
| Any combination | Dichotomous (e.g., yes/no) | Binary/Binomial Logit Model |
| Any combination | Polytomous, Unordered (e.g., vote) | Multinomial Logit Model, Nested Dichotomies? |
| Any combination | Polytomous, Ordered (e.g., education categories) | Proportional-Odds Model?, Continuation Dichotomies? |
| Any combination | Count | Poisson/Quasi-Poisson Generalized Linear Model |