# Regression Models for Qualitative/Quantitative Predictors and ANOVA 

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## Outline

Quantitative Predictors

Qualitative Predictors

Why Qualitative Predictors?

ANOVA vs. Regression

Chapter 8, 16.

## Variations on regression models

- Polynomial regression models: e.g.

$$
\begin{aligned}
Y_{i} & =\beta_{0}+\beta_{1} x_{i}^{\prime}+\beta_{2} x_{i}^{\prime 2}+\epsilon_{i}, i=1, \cdot, n . \text { where } x_{i}^{\prime}=x_{i}-\bar{x} \text { or } \\
x_{i}^{\prime} & =\left(x_{i}-\bar{x}\right) / s d(x) \text { etc. }
\end{aligned}
$$

- Interaction Regression models: e.g. $E(Y \mid x)=\beta_{0}+\beta_{1} x_{1}+\beta_{2} x_{2}+\beta_{3} x_{1} x_{2}$. Graphical Illustrations.
- General form: $E(Y \mid x)=f_{1}\left(x_{1}\right)+f_{2}\left(x_{2}\right)+f_{3}\left(x_{1}, x_{2}\right)$
- Interpretation of parameters
- Numerical stable, practically interpretable and flexible


# Interaction with indicator 

Two simple linear regression models with the normal errors with common variance.

## Interaction with indicator

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$$
\begin{array}{ll}
Y_{i}=a_{1}+b_{1} X_{i}+\epsilon_{i}, & i=1, \cdots, n \\
Y_{j}=a_{2}+b_{2} X_{j}+\epsilon_{j}, & j=n+1, \cdots, n+m \tag{2}
\end{array}
$$

Trend changes

## Interaction with indicator

Two simple linear regression models with the normal errors with common variance.

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\end{array}
$$

Trend changes

$$
\begin{array}{ll}
Y=a_{1}+b_{1} X+\epsilon, & \text { for } X<x_{0} \\
Y=a_{2}+b_{2} X+\epsilon, & \text { for } X \geq x_{0}
\end{array}
$$

Some problems call for alternative models than regression.

- Which grad school is the best?
- Which treatment is better? (Program A, B, C)
- What dosage level (low, medium, high) is most effective?
- What is the best treatment combination to manufacture a product?


## Math form

- $E(Y \mid X)=\beta_{0}+\beta_{1} X_{1}+\beta_{2} X_{2}$
- $E(Y \mid \tau, B)=\mu+\tau+B$


## Qualitative vs. Quantitative

- Picture (Figure 16.1, KNNL)
- Factor, Factor Level ("Value" of the factor)
- Spectrum from Quantitative-Qualitative variables. Categorical Variables.
- Single factor versus Multifactor


## Single Factor ANOVA

- Cell Means Model

$$
Y_{i j}=\mu_{i}+\epsilon_{i j}, i=1, \cdots, r ; j=1, \cdots, n_{i}
$$

$$
\epsilon_{i j} \sim_{i i d} N\left(0, \sigma^{2}\right)
$$

- Factor Effects Model

$$
\begin{aligned}
& Y_{i j}=\mu+\tau_{i}+\epsilon_{i j}, i=1, \cdots, r ; j=1, \cdots, n_{i}, \\
& \epsilon_{i j} \sim \sim_{i i d} N\left(0, \sigma^{2}\right)
\end{aligned}
$$

- Connection

$$
\mu .=\sum_{i=1}^{r} \mu_{i} / r ; \sum \tau_{i}=0
$$

- Both models are GLM satisfying GM conditions.

Connection with two-sample $t$-tests

## Regression vs. ANOVA

- Design Matrices for GLM
- Typical Question of interest: $\beta=, \geq, \leq 0$ ? vs $\sum_{i} \tau_{i}^{2}=0$ ? and ordering in $\tau$.
- Calculation glm vs. $\underline{\mathrm{Im}}$

Choice of models

- Quantitative predictor: Resolution, Precision vs. Robustness
- Qualitative predictor: type of ordering. scoring.

