

An aerial photograph of a city with a large river in the foreground. In the middle ground, there are several large, multi-story buildings, likely university buildings, surrounded by greenery. The background shows a dense urban area with many smaller buildings and a hazy sky.

A Sampling of Data Education in ChE Curricula

**Presented by Richard D. Braatz
(with input from others)**



Outline

- **Data education in undergraduate ChE curricula (a sampling)**
- **Data education in a graduate ChE curriculum**

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Data Education in ChE Curricula

- **Data education ranges from**
 - **A few lectures in some chemical engineering course(s)**
 - **3.5 weeks in a chemical engineering course**
 - **Statistics and probability course taught by statistics/math faculty**
 - **Engineering statistics taught by a non-ChE engineer**
 - **Engineering statistics course taught by ChE faculty**
- **MIT is rare in that all of its graduate students take some statistics, which is covered in two courses, and the training includes more advanced methods**

Data Education in Undergraduate ChE Curricula

- **Universities sampled**
 - **University at Buffalo**
 - **University of Texas Austin**
 - **University of Massachusetts, Amherst**
 - **Massachusetts Institute of Technology**
 - **Brigham Young University**
- **A good coverage of different amounts and approaches used in ChE curricula**

Data Education in Undergraduate ChE Curricula

- **University at Buffalo, 14 weeks to juniors**
- **Lecturer: David A. Kofke (ChE)**
- **William Navidi, *Statistics for Engineers & Scientists***
- **Sampling and descriptive statistics, probability, error propagation, common distributions, confidence intervals, hypothesis testing, factorial experiments**

Data Education in Undergraduate ChE Curricula

- **University of Texas Austin, 16 weeks to juniors**
- **Lecturer: Keith Friedman (ChE)**
- **R.A. Johnson, *Statistics & Probability for Engineers***
- **Linear regression, JMP, simple distributions, confidence intervals, hypothesis testing, ANOVA, design of experiments, statistical process control**
- **Taught by ChE lecturer**

Data Education in Undergraduate ChE Curricula

- **University of Massachusetts, 3.5 weeks to juniors**
- **Lecturer: Michael A. Henson**
- **Erwin Kreyszig, *Advanced Engineering Mathematics***
- **Probability distributions, confidence intervals, hypothesis testing, regression and correlation, factorial and fractional factorial experimental design, Matlab statistics**

Data Education in Undergraduate ChE Curricula

- **Massachusetts Inst. Tech., small number of lectures to seniors in design and project courses**
- **Lecturers: numerous**
- **No textbooks**
- **Laboratory kinetic data and curve fitting**

Data Education in Undergraduate ChE Curricula

- Brigham Young University, 14 weeks to sophomores, 14 weeks elective for seniors and graduate students
- Lecturers: Matt Heiner (Stats), Larry Baxter (ChE), John Hedengren (ChE)
- William Navidi, *Statistics for Engineers & Scientists, Lecture Notes and Online for Seniors/Grad Students*
- The scientific method; probability, random variables, common discrete and continuous random variables, central limit theorem; confidence intervals and hypothesis testing; completely randomized experiments; factorial experiments

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- Data education in undergraduate ChE curricula (a sampling)
- Data education in a graduate ChE curriculum

Data Education in a Graduate ChE Curriculum

- **MIT, 3 weeks (9 hours) to all graduate students**
- **Lecturers: Richard D. Braatz and James W. Swan**
- **Electronic lecture notes**
- **Probability theory, stochastic differential equations, parameter estimation, Monte Carlo methods, stochastic chemical kinetics**
- **Clear that most entering students do not have a basic understanding of probability and statistics**

Data Education in a Graduate ChE Curriculum

- MIT, 3.5 weeks (10 hours) to most graduate students
- Lecturer: Richard D. Braatz
- Electronic lecture notes
- Statistical and model-based iterative experimental design, linear and nonlinear regression (parameter estimation), uncertainty quantification, control charts, chemometrics for sensor calibration and process monitoring, machine learning for construction of sparse models

Data Education in a Graduate ChE Curriculum

- **Main goal: train students to be effective in translating data into making good decisions**
 - **Experimental design** \Rightarrow generate data so that the model will be good enough
 - **Linear/nonlinear regression** \Rightarrow models for design & control
 - **Uncertainty quantification** \Rightarrow is the model good enough?
 - **Chemometrics** \Rightarrow handling correlated data

Data Education in a Graduate ChE Curriculum

- **Main goal: train students to be effective in translating data into making good decisions**
 - **Statistical process control**
 - ⇒ does data indicate that the process is under control?
 - ⇒ which variables are likely associated with the fault?
 - ⇒ how do classify new data based on historical data
 - **Chemometrics (i.e., principal component analysis, partial least squares) and Fisher discriminant analysis**
 - **Machine learning for construction of sparse models, e.g., sparse vs. dense models, lasso & elastic net methods**

Data Education in a Graduate ChE Curriculum: Experimental Design

- **Major steps for designing & carrying out a study:**
 1. **State objectives, assumptions, hypotheses**
 2. **Draw up preliminary design: materials, procedures, ...**
 3. **Review with collaborators, e.g., assess potential biases**
 4. **Draw up final design, including data analysis methods**
 5. **Carry out design: record data, record modifications**
 6. **Analyze data: review, graph, apply data analysis methods**
 7. **Interpret results: confine to evidence, assess significance**
 8. **Write report: background, tables/figures, limit to evidence**

Data Education in a Graduate ChE Curriculum: Sensor Calibration, Regression, Uncertainty Quantification

- Start with relating spectra to concentration
- Do linear and nonlinear least squares for constructing algebraic sensor calibration curves, using summation notation and matrix algebra
- Statistical process control: Shewart, CUSUM, EWMA, PCA-based T^2 , 1D/2D contribution plots
- Do chemometrics for handling correlated data
- Do parameter estimation for nonlinear dynamic models, quantify uncertainties in parameters