

ECE 307 – Techniques for Engineering Decisions

13. Data Uses

George Gross

**Department of Electrical and Computer Engineering
University of Illinois at Urbana–Champaign**

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

1

FOCUS OF DATA USAGE TOPIC

- Use of historical data for the construction of probability distributions**
- The interpretation of probability information**
- Use of estimators**
- Application example**

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

2

EXAMPLE

- Consider the interpretation of the statement

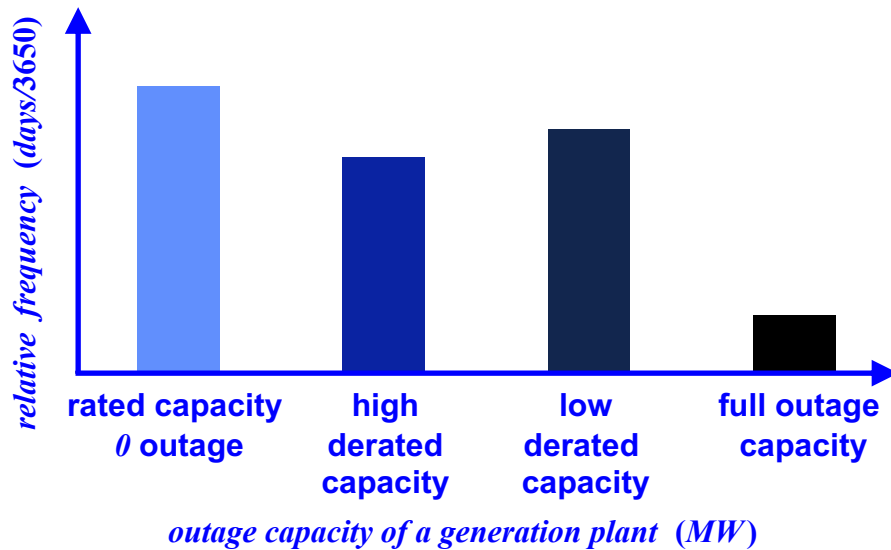
$$P\{\textit{sunny day in June in Champaign}\} = 0.53$$

- We obtain this probability from, say, 20 years of June weather data in Champaign, with each day classified as either *sunny* or *not sunny*
- The 600 June days of data indicate that 318 or 53 % of these days are classified as sunny
- Given the long-term historical behavior in the data, the probability of 0.53 makes sense

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

3

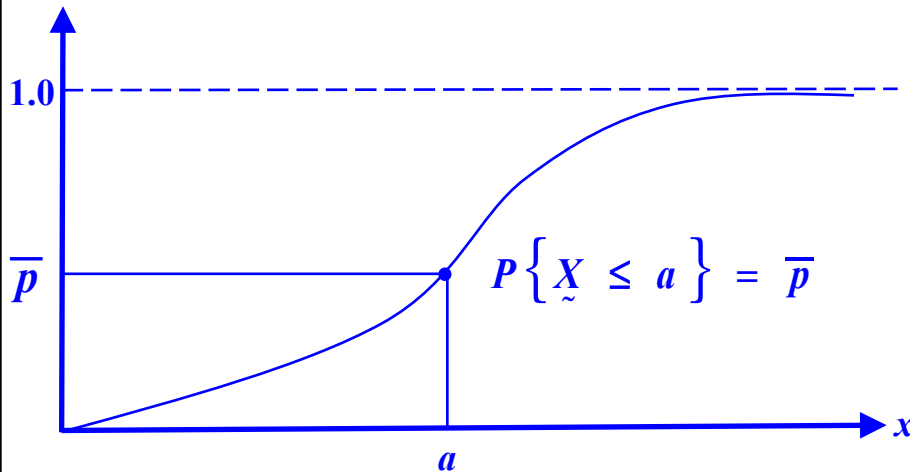
USE OF HISTOGRAMS



© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

4

CONSTRUCTION OF THE *c.d.f.*



© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

5

STATISTICAL PARAMETER ESTIMATORS

- An estimator is a *r.v.* that can be used to estimate the value of a parameter of interest
- Consider a *r.v.* \tilde{X} whose **statistical parameters** we wish to estimate
- We consider a set of *r.v.s* $\{\tilde{X}_i, i = 1, 2, \dots, n\}$, where each \tilde{X}_i is independent of $\tilde{X}_j, i \neq j$, and each \tilde{X}_i has the same distribution as \tilde{X} ; we refer to this set as a set of n **independent, identically distributed** or ***i.i.d.*** *r.v.s*

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

6

STATISTICAL PARAMETER ESTIMATORS

□ We use the set of n *i.i.d.* *r.v.s* $\{X_{\tilde{i}}, i = 1, 2, \dots, n\}$ to construct estimators for the moments of $X_{\tilde{}}$

□ We focus on the estimators for two key parameters of $X_{\tilde{}}$:

○ the mean of $X_{\tilde{}}$

○ the variance of $X_{\tilde{}}$

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

7

STATISTICAL PARAMETER ESTIMATORS

□ The *sample mean estimator* is the *r.v.*

$$\bar{X}_{\tilde{}} = \frac{\sum_{i=1}^n X_{\tilde{i}}}{n}$$

□ In practice, we obtain an estimate of the mean by using the *observed realizations* of the n *r.v.s* $X_{\tilde{i}}$

$$\bar{x} = \frac{\sum_{i=1}^n x_i}{n}$$

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

8

STATISTICAL PARAMETER ESTIMATORS

- The estimator of the sample variance is given by the *r.v.*

$$\underline{S}^2 = \frac{\sum_{i=1}^n (\underline{X}_i - E\{\underline{X}\})^2}{n-1}$$

- We obtain an estimate of the variance by using the observed realizations of the n *r.v.s* \underline{X}_i

$$s^2 = \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n-1}$$

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

9

STATISTICAL PARAMETER ESTIMATORS

- An equivalent way to think about the computation of the estimate is to draw n random samples from the sample space of \underline{X}

- We collect the set of n random samples

$\{x_1, x_2, \dots, x_n\}$ of the *r.v.* \underline{X} : these are n *randomly*

drawn values from the sample space of \underline{X}

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

10

STATISTICAL PARAMETER ESTIMATORS

- The value \bar{x} computed with the set of random samples provides an estimate of

$$\mu = E\{X\}$$

- The value s^2 computed with the set of random samples provides an estimate of

$$\sigma^2 = var\{X\}$$

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

11

EXAMPLE: TACO SHELLS

- This application example focuses on taco shells and is concerned with the high breakage rate in the shipment of most taco shells: typical rate is 10 – 15 %
- A company with a new shipping container claims to have a lower – approximately 5% – *breakage rate*
- This company's price is \$ 25 for a 500-taco shell box vs. \$ 23.75 for a 500-taco shell box of the current supplier

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

12

EXAMPLE: TACO SHELLS

- A test run using 12 boxes from the new company and 18 boxes from the current company is performed and used for comparison purposes: we randomly pick the elements to construct the set

$$\{x_1, x_2, \dots, x_{12}\}$$

from the sample space of the *r.v.* \tilde{X} to represent

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

13

EXAMPLE: TACO SHELLS

the number of unbroken shells from the new company and the elements to construct the set

$$\{y_1, y_2, \dots, y_{18}\}$$

from the sample space of the *r.v.* \tilde{Y} to represent those of the current company

- We tabulate the data of the useable shells from the two suppliers

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

14

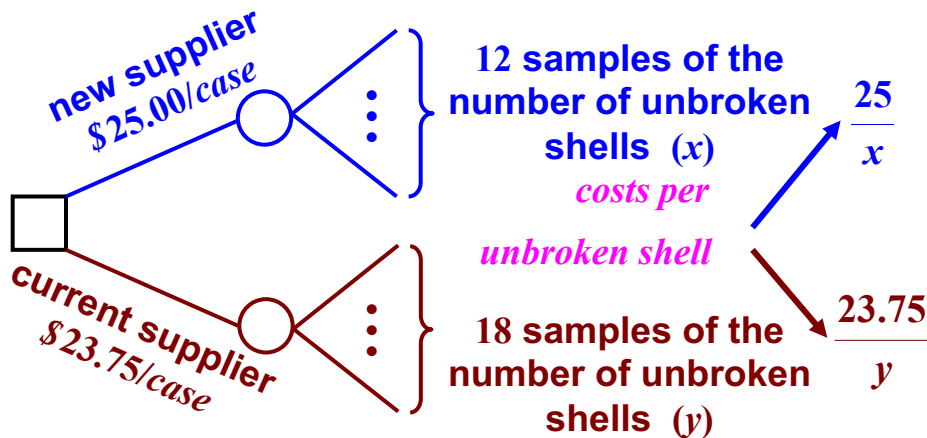
UNBROKEN TACO SHELLS

<i>new supplier</i>		<i>current supplier</i>		
468	467	444	441	450
474	469	449	434	444
474	484	443	427	433
479	470	440	446	441
482	463	439	452	436
478	468	448	442	429

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

15

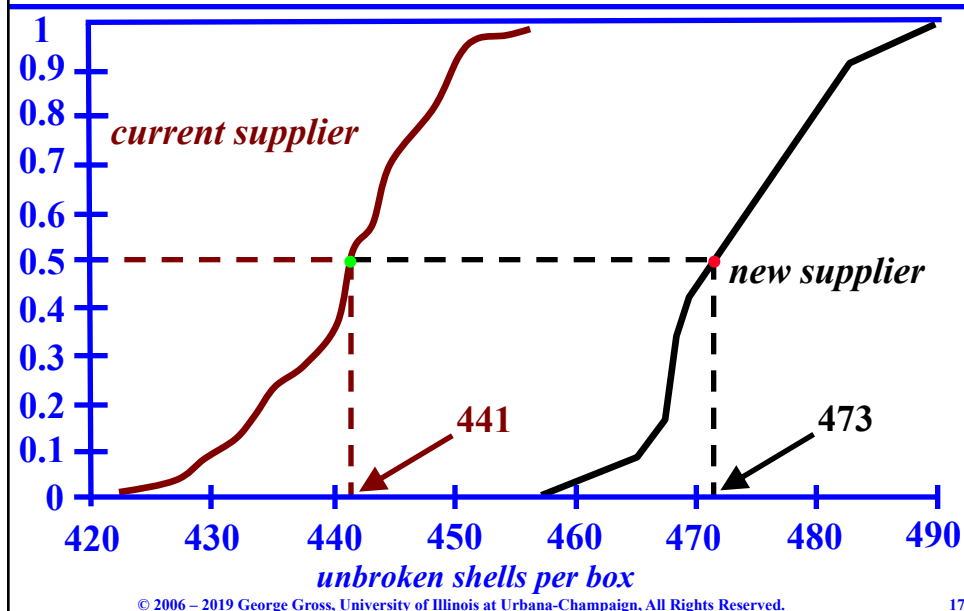
EXAMPLE: TACO SHELLS



© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

16

c.d.f.s CONSTRUCTED FOR THE TWO SUPPLIERS



© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

17

c.d.f.s OF THE TWO SUPPLIERS

- Clearly, the new supplier has the higher expected number of useable shells per box; the two distributions, however, are highly similar
- The mean number of useable shells for the new supplier is 473 and so the expected costs per

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

18

c.d.f.s OF THE TWO SUPPLIERS

useable shell is \$0.0529; the minimum (maximum)

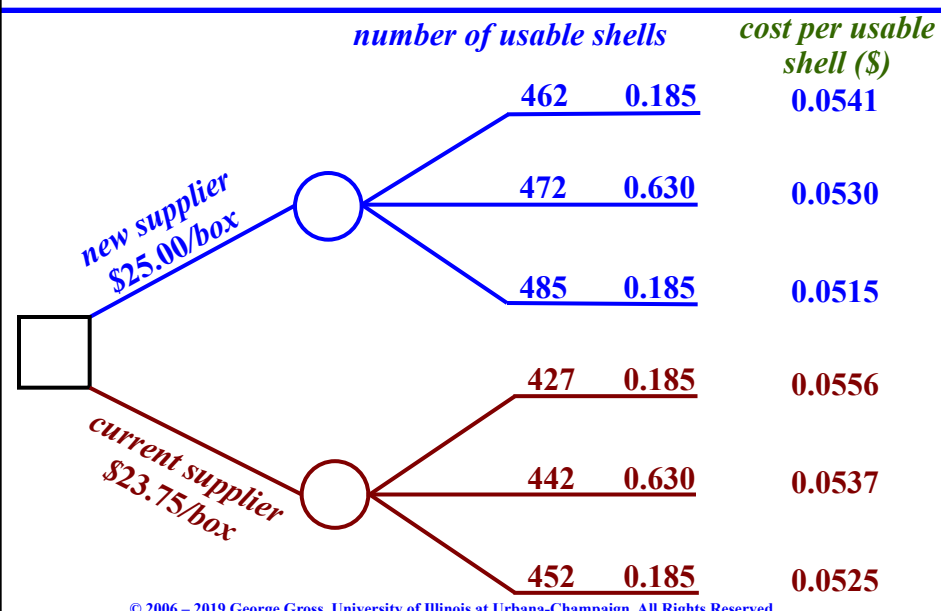
number of useable shells is 463(482)

- The mean number of useable shells for the current supplier is **441** and so the expected costs per useable shell is **\$0.0539**; the minimum (maximum) number of useable shells is **429(452)**

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

19

REDUCED ORDER REPRESENTATION OF THE TEST RUN DATA



© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

20

COMMENTS

- We use the observed sample-based *c.d.f.s* to estimate the mean of each supplier population

- In general for an arbitrary *r.v.* \tilde{X}

$$E\left\{\frac{1}{\tilde{X}}\right\} \neq [E\{\tilde{X}\}]^{-1}$$

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

21

COMMENTS

and so we cannot use the approximation

$$E\left\{\frac{25}{\tilde{X}}\right\} \approx \frac{25}{E\{\tilde{X}\}}$$

- This example demonstrates the usefulness of the *c.d.f.s* in applications even when they can only be approximated for the limited data available

© 2006 – 2019 George Gross, University of Illinois at Urbana-Champaign, All Rights Reserved.

22