
ECE 307 – Techniques for Engineering Decisions

13. Data Uses

George Gross

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

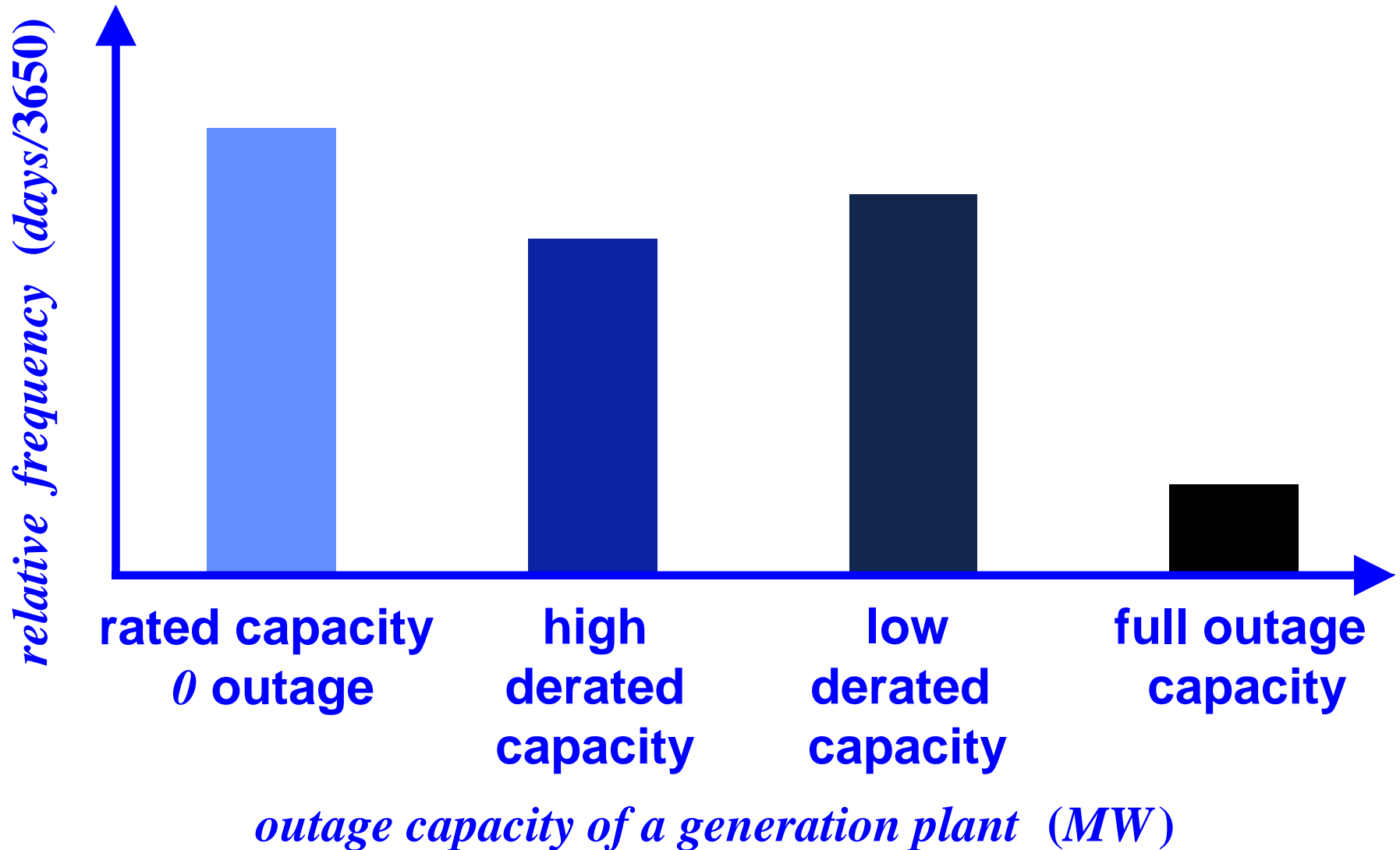
FOCUS OF DATA USAGE TOPIC

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

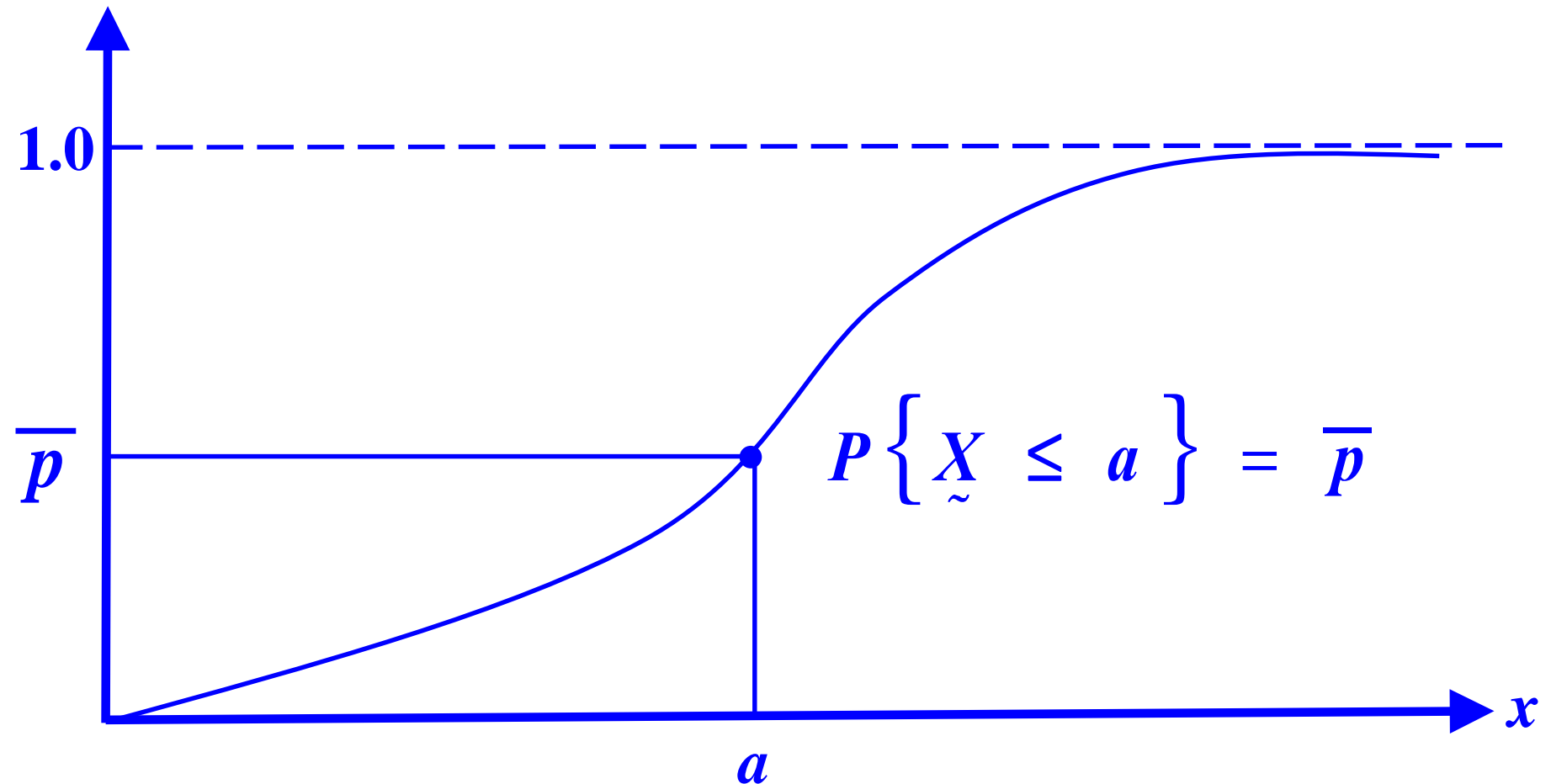
EXAMPLE

- ❑ Consider the interpretation of the statement
$$P\{ \textit{sunny day in June in Champaign} \} = 0.53$$
- ❑ We obtain this probability from 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

USE OF HISTOGRAMS



CONSTRUCTION OF THE *c.d.f.*



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*

STATISTICAL PARAMETER ESTIMATORS

- We use the set of n *i.i.d. r.v.s* $\{X_{\sim i}, i = 1, 2, \dots, n\}$ to construct estimators for the moments of X_{\sim}
- We focus on the estimators for two key parameters of X_{\sim} :
 - mean of X_{\sim}
 - variance of X_{\sim}

STATISTICAL PARAMETER ESTIMATORS

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

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

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

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

STATISTICAL PARAMETER ESTIMATORS

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

$$s_{\sim}^2 = \frac{\sum_{i=1}^n \left(X_{\sim i} - E\{X_{\sim}\} \right)^2}{n-1}$$

- We obtain an estimate of the variance by using the observed realizations of the n *r.v.s* $X_{\sim i}$

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

STATISTICAL PARAMETER ESTIMATORS

- An equivalent way to think about the computation of the estimate is to draw n random samples from the sample space \underline{X}
- We collect the set of n random samples $\left\{x_1, x_2, \dots, x_n\right\}$ of the *r.v.* \underline{X} : these are n *randomly* drawn values from the sample space of \underline{X}

STATISTICAL PARAMETER ESTIMATORS

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

$$\mu = E\{X_{\sim}\}$$

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

$$\sigma^2 = \text{var}\{X_{\sim}\}$$

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

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

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

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

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.* \underline{Y}_{\sim} to represent those of the current company

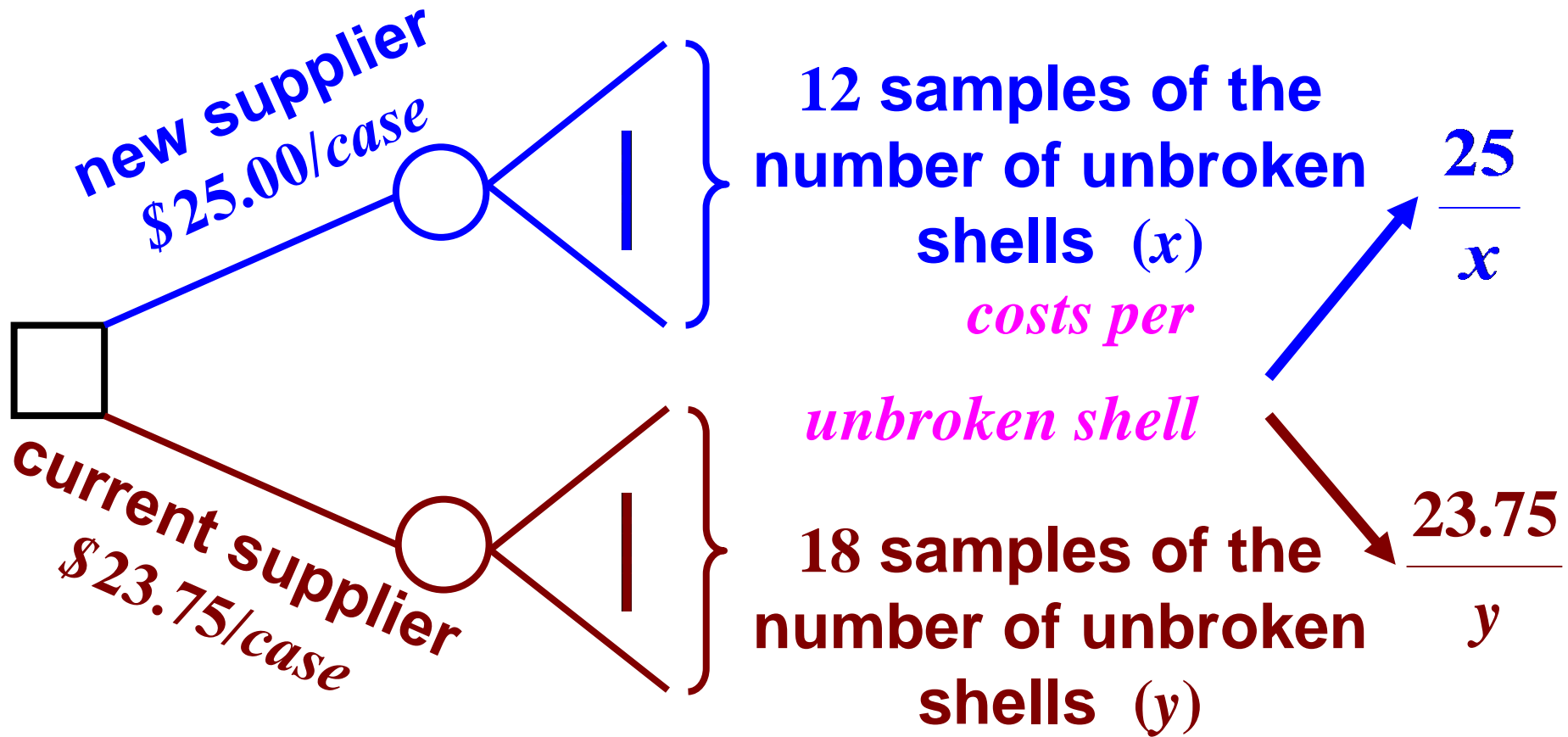
- We tabulate the data of the useable shells from the two suppliers

UNBROKEN TACO SHELLS

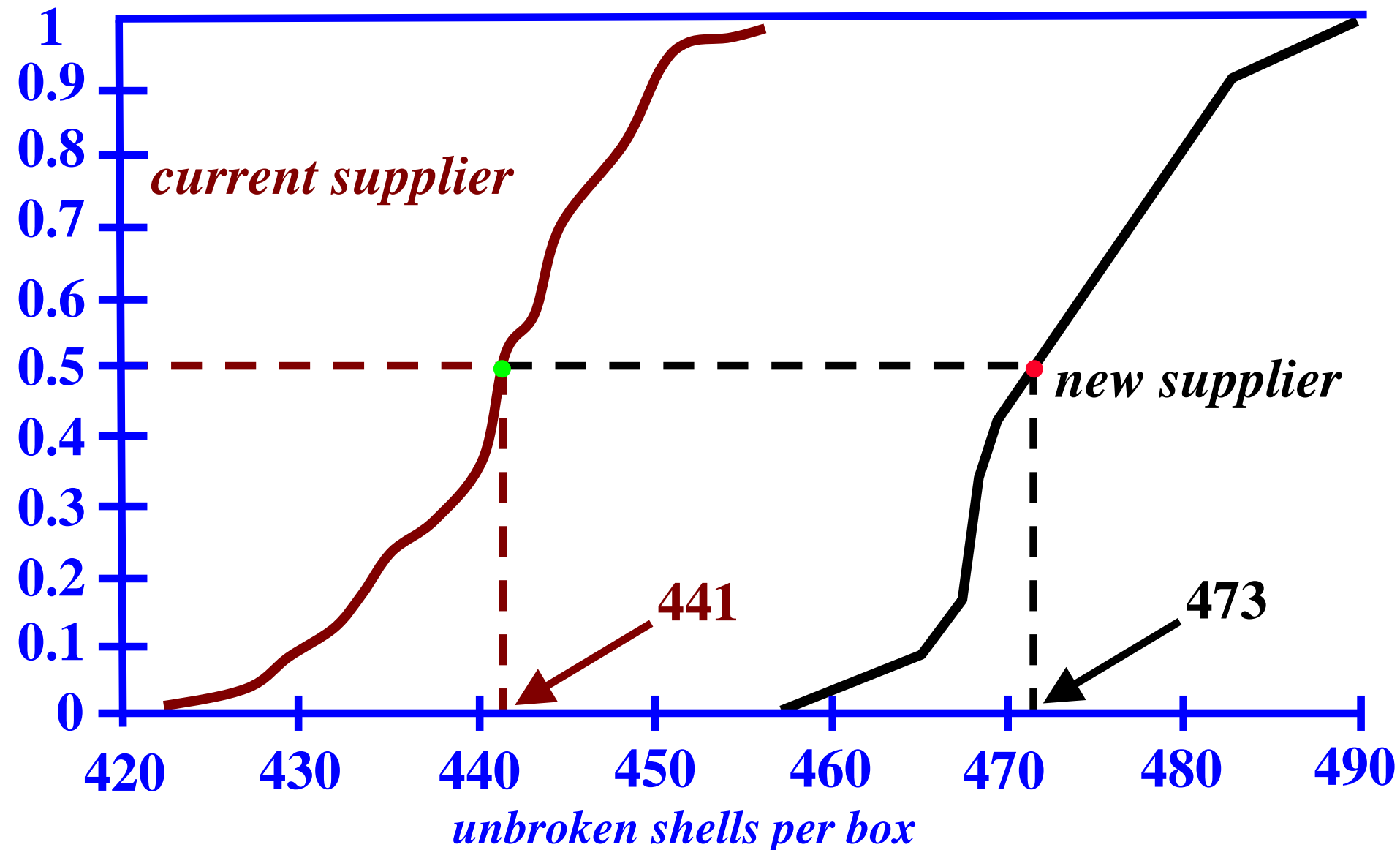
<i>new supplier</i>	
468	467
474	469
474	484
479	470
482	463
478	468

<i>current supplier</i>		
444	441	450
449	434	444
443	427	433
440	446	441
439	452	436
448	442	429

EXAMPLE: TACO SHELLS



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



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

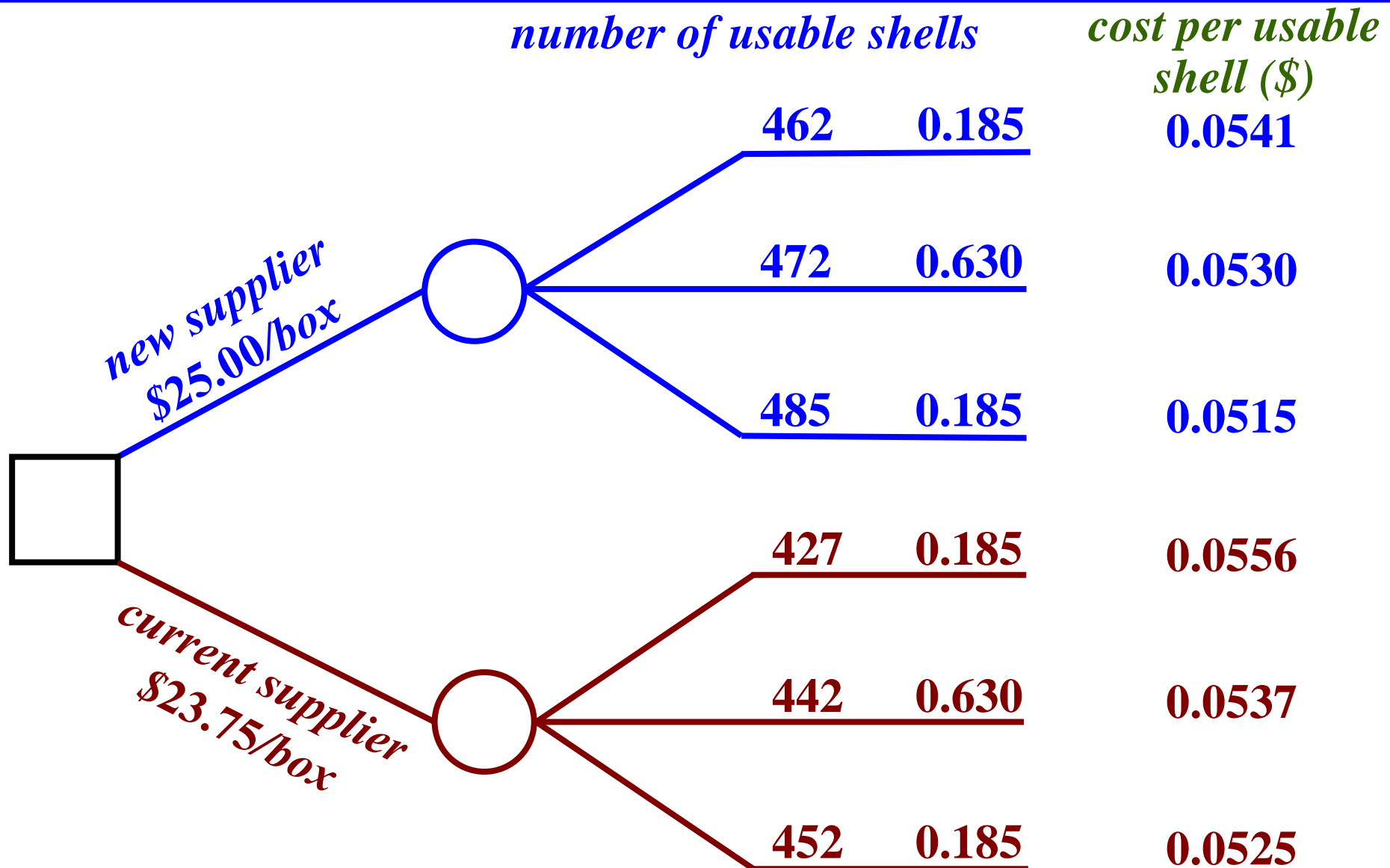
c.d.f.s OF THE TWO SUPPLIERS

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)

REDUCED ORDER REPRESENTATION OF THE TEST RUN DATA



COMMENTS

□ We use the *c.d.f.s* to estimate the means of the

two populations of suppliers

□ In general for an arbitrary *r.v.* \underline{X}_{\sim}

$$E\left\{\frac{1}{\underline{X}_{\sim}}\right\} \neq \left[E\left\{\underline{X}_{\sim}\right\}\right]^{-1}$$

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