

# Probability Plots

Keegan Korthauer  
Department of Statistics  
UW Madison

# Recap

## Summary Statistics

- Measuring central tendency
  - Mean
  - Median
  - Mode
- Measuring spread
  - Standard deviation
  - Percentiles
- Graphical summary
  - Histogram
  - Box plot

## Probability

- Conditional probability
  - Total probability
  - Bayes rule
- Distributions
  - Discrete: Binomial, Poisson, Geometric
  - Continuous: Normal, Uniform, Exponential

## Statistical Inference

- Point estimation
- Central Limit Theorem
- Confidence intervals
- Hypothesis testing
- Simple Linear Regression
- Multiple Regression

# PROBABILITY PLOTS

How to construct  
Interpretation

# How to Choose a Distribution?

- So far we've considered two scenarios:
  1. We know what distribution our data follow and are given the parameter values
  2. We know that our data come from a certain distribution but do not know the parameter values so we estimate them (e.g.  $\hat{p} = X / n$  in the binomial case) and their uncertainty
- A third scenario to consider: we *suspect* that our data follow a certain distribution, but we don't know for sure.  
**How do we check?**

# Comparing Sample Distributions to Population Distributions

- Say we have a sample of 5 measurements that we suspect come from a normal distribution:

3.01, 3.35, 4.79, 5.96, 7.89

- Compare the distribution of our sample with the distribution of the suspected population (normal, in this case) to see if they are similar using a **probability plot!**

# Intuition Behind the Probability Plot

- For each original observation  $X_i$  in our sample, find its percentile
- For each of these percentiles, find the quantile  $Q_i$  of the suspected distribution that corresponds to it
- Examine the ordered pairs  $(X_i, Q_i)$ 
  - If the data do come from the suspected distribution, they will lie close to a straight line
  - If they come from some other distribution, the points could be far from a straight line

# How to Construct a Probability Plot

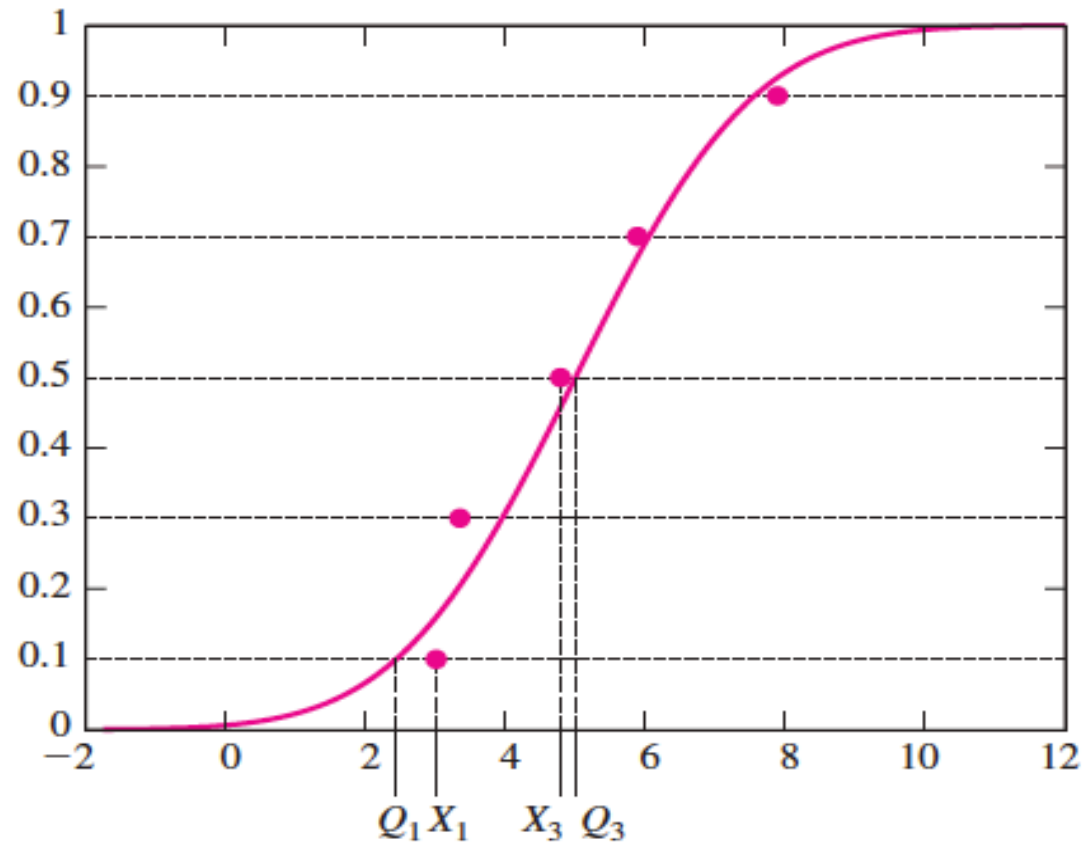
- First find the sample ‘percentiles’
- For each of these find the quantile of the suspected normal distribution
  - for  $i=1$ , we find  $Q_1$  such that  $P(X \leq Q_1) = 0.1$  when  $X \sim N(\mu, \sigma^2)$
  - best guess for  $\mu = 5$  and  $\sigma = 2$  (sample mean and standard deviation)
  - Standardize:  $P(Z \leq (Q_1 - 5)/2) = 0.1$
  - From table:  $P(Z \leq -1.28) \approx 0.1$
  - So  $Q_1 = 2 * (-1.28) + 5 = 2.44$

$i$	$X_i$	“Percentiles” $(i-0.5)/n$	$Q_i$
1	3.01	0.1	2.44
2	3.35	0.3	3.95
3	4.79	0.5	5.00
4	5.96	0.7	6.05
5	7.89	0.9	7.56

Not true percentiles,  
but evenly spaced  
from 0 to 1

Plot  $X_i$  vs  $Q_i$

# Visualizing $Q_i$

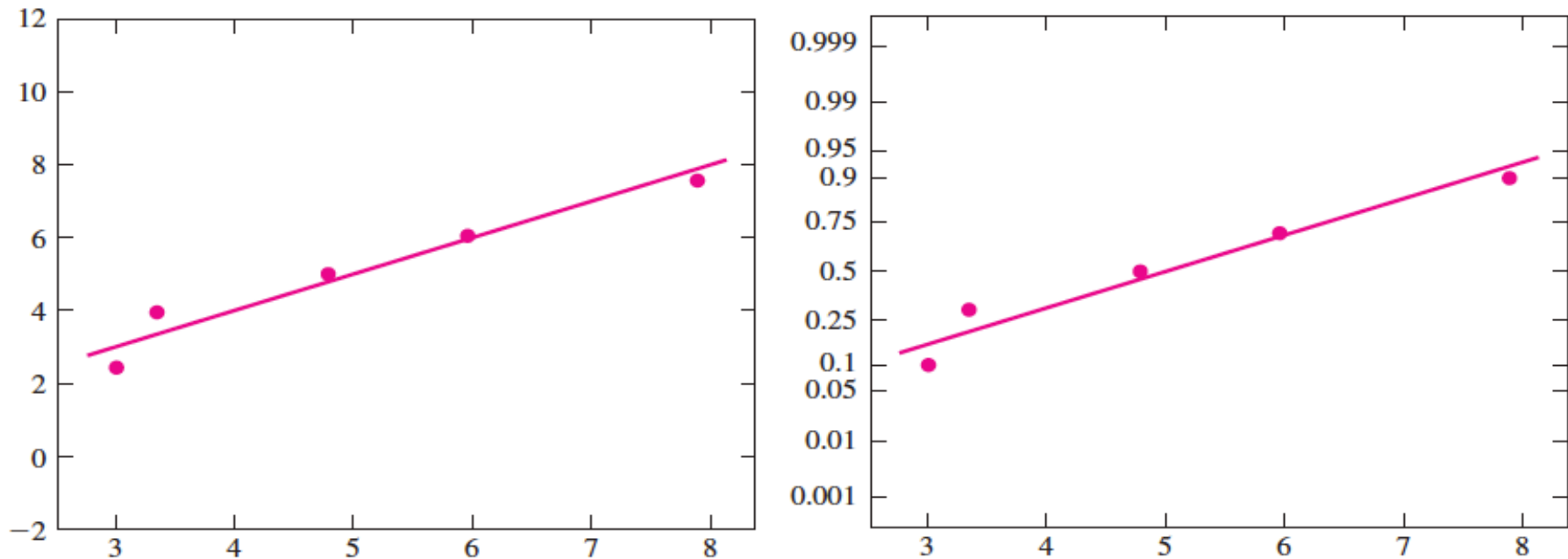


**FIGURE 4.21** The curve is the cdf of  $N(5, 2^2)$ . If the sample points  $X_1, \dots, X_5$  came from this distribution, they are likely to lie close to the curve.



# Probability Plots

The **probability plot** consists of the points  $(X_i, Q_i)$ . Since the distribution that generated the  $Q_i$  was a normal distribution, this is called a **normal probability plot**.

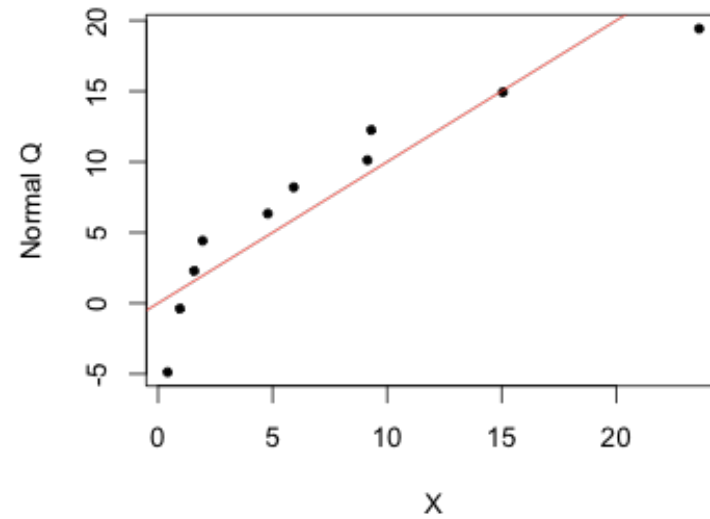
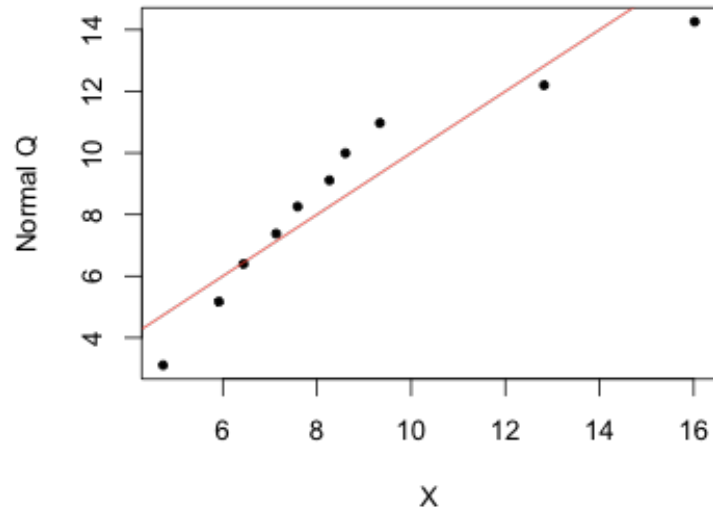


**FIGURE 4.22** Normal probability plots for the sample  $X_1, \dots, X_5$ . The plots are identical, except for the scaling on the vertical axis. The sample points lie approximately on a straight line, so it is plausible that they came from a normal population.

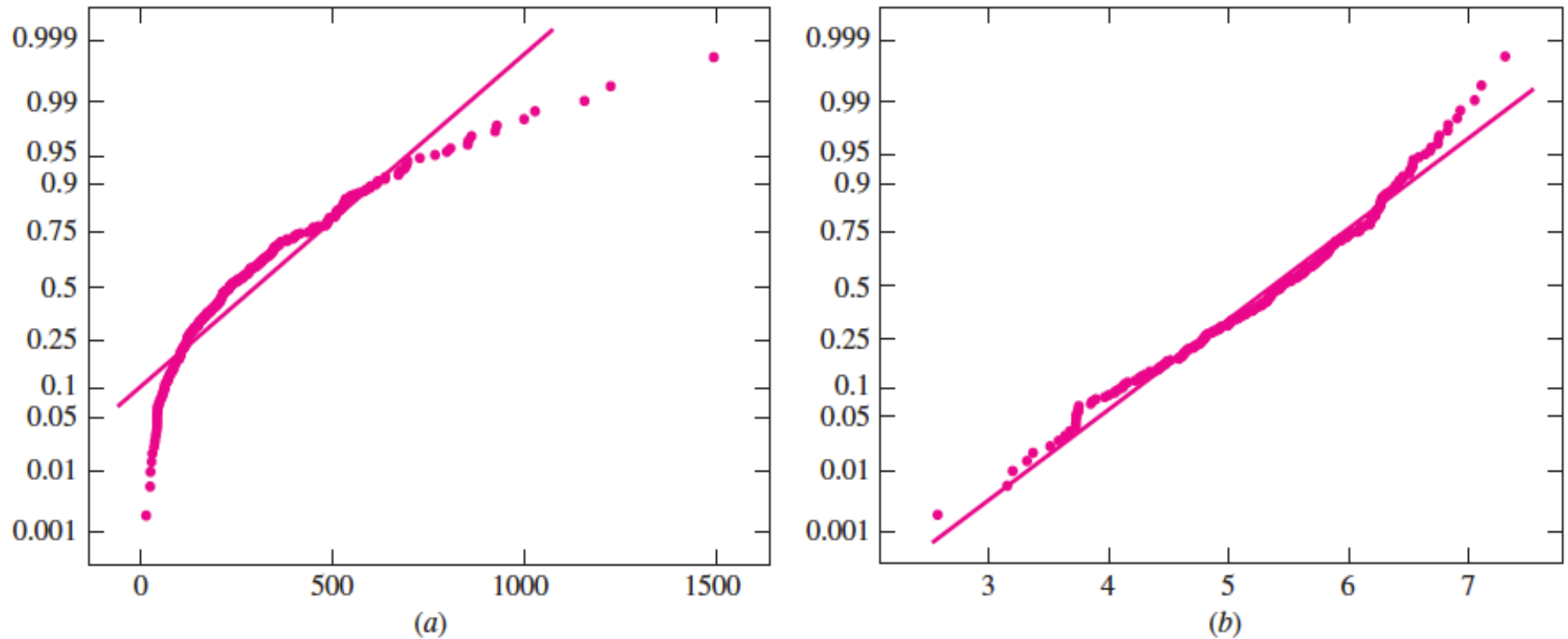
# Sample Size

- With a small sample, probability plots will only show large departures from the suspected distribution
- Rule of thumb: sample size of **30 or more** will yield a reliable probability plot
- Use computer program (like R) to generate plots

# Example – Sample of Size 10 vs 500



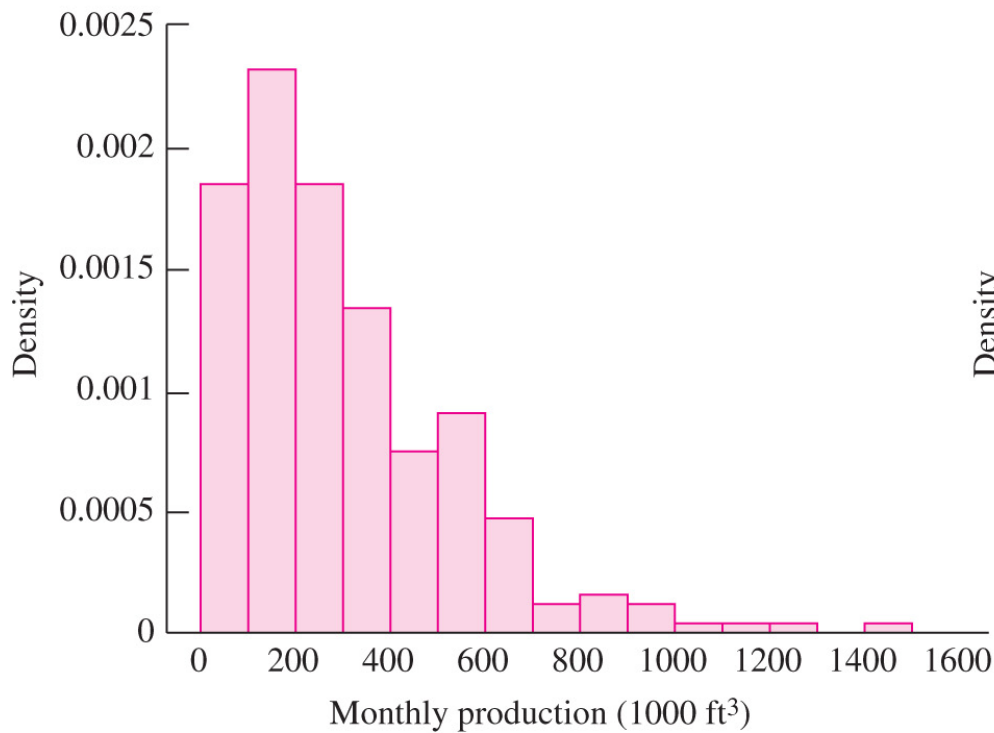
# Example - Normal Probability Plots



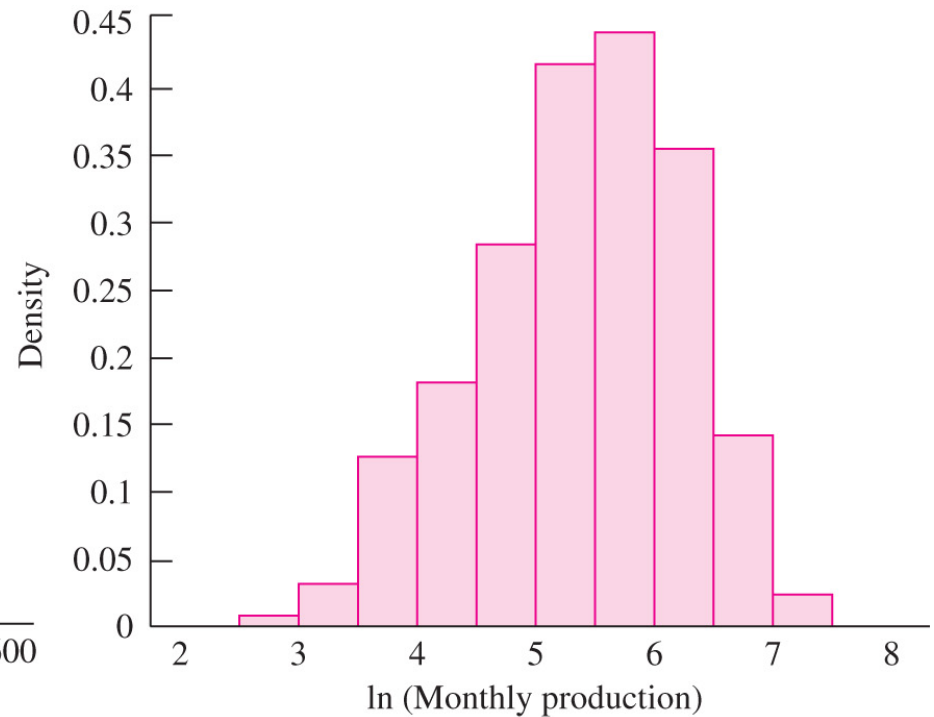
**FIGURE 4.23** Two normal probability plots. **(a)** A plot of the monthly productions of 255 gas wells. These data do not lie close to a straight line, and thus do not come from a population that is close to normal. **(b)** A plot of the natural logs of the monthly productions. These data lie much closer to a straight line, although some departure from normality can be detected. See Figure 4.16 for histograms of these data.

# Example - Normal Probability Plots

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(a)



(b)

# Interpretation of Probability Plots

- No hard-and-fast rules – use the ‘eye-ball’ method
- Look for strong trends
- Common for a few points at the ends to stray
- Outliers will be far from the line when most of the others are close

# Now What?

- Your plot shows strong departure from your suspected distribution. So what can you do?
  - Try plotting against the quantiles of a different distribution
  - Transform your data – more on this in Chapter 7
    - log-transformation
    - square root transformation
    - power transformation

# Next

- Central Limit Theorem
- Introduction to R
- Exam 1 Review