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The log-normal distribution in R

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A log-normal distribution is completely determined by two parameters, μ and σ , the mean and standard deviation of the corresponding normal distribution. These are **not** the mean and standard deviation of the log-normal distribution itself!



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The following code gives 12 values from a distribution whose logarithm is normal with mean 6.1 and standard deviation 1.0.

```
## [1] 1576.48708 321.74736 1685.46920 1591.49534 674.95190 95.58826
```

rlnorm(12, 6.1, 1.0)

[7] 176.16710 332.04801 443.29383 4937.69247 956.79912 200.53539



• plnorm(q, μ , σ) returns the probability that a random observation X is less than or equal to q. It's the cumulative distribution function of the log-normal, $P(X \leq q)$.



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• $plnorm(q, \mu, \sigma)$ returns the probability that a random observation X is less than or equal to q. It's the cumulative distribution function of the log-normal, $P(X \leq q)$. As usual in R, the quantity q can be a vector, so you can compute several probabilities at once.

The following code computes $P(X \le 200)$, $P(X \le 400)$, $P(X \le 800)$, and $P(X \le 1600)$ in the distribution whose logarithm is normal with mean 6.1 and standard deviation 1.0.

[1] 0.2113683 0.4567855 0.7205956 0.8993328



• $qlnorm(p, \mu, \sigma)$ returns the quantity q such that $plnorm(q, \mu, \sigma) = p$. It's the inverse cdf of the specified log-normal distribution. Again, p can be a vector.



• qlnorm(p, μ , σ) returns the quantity q such that plnorm(q, μ , σ) = p. It's the inverse cdf of the specified log-normal distribution. Again, p can be a vector.

The following code computes the 10^{th} , 30^{th} , 50^{th} , 70^{th} , and 90^{th} percentiles in the distribution whose logarithm is normal with mean 6.1 and standard deviation 1.0.

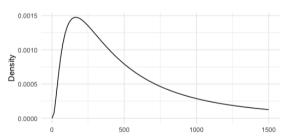
```
qlnorm(c(.1, .3, .5, .7, .9), 6.1, 1.0)
## [1] 123.7729 263.9077 445.8578 753.2525 1606.0798
```



• dlnorm(x, μ , σ) is the probability density function (pdf) for the specified log-normal distribution.



• dlnorm(x, μ , σ) is the probability density function (pdf) for the specified log-normal distribution. It's mostly used for graphing and for theoretical calculations.





Example. Freshman undergraduate enrollments at U.S. colleges have an approximate log-normal distribution with parameters $\mu = 6.1$ and $\sigma = 1.0$.

1. What is the probability that a randomly-selected college enrolls between 500 and 1000 freshmen in a year?

2. What is the 99th percentile for U.S. college enrollment?

3. Simulate selecting 100 colleges at random. Plot a histogram of freshman enrollments at those colleges.

