

T - Distribution

Egor Howell

What is the t-distribution?

The [t-distribution](#), is a continuous probability distribution that is very similar to the [normal distribution](#), however has the following key differences:

- **Heavier tails:** *More of its probability mass is located at the extremes (higher [kurtosis](#)). This means that it is more likely to produce values far from its mean.*
- **One parameter:** *The t-distribution has only one parameter, the [degrees of freedom](#), as it's used when we are unaware of the population's variance.*

Origin

The origin behind the t-distribution comes from the idea of modelling normally distributed data without knowing the population's variance of that data.

For example, say we sample n data points from a normal distribution, the following will be the mean and variance of this sample respectively:

$$\bar{x} = \frac{x_1 + \dots + x_n}{n}$$

Equation by author in LaTeX.

Sample Mean

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

Equation by author in LaTeX.

Sample Standard Deviation

Origin

Combining the two equations, we can construct the following random variable:

The diagram illustrates the construction of a t-statistic. It features three equations arranged vertically. The top equation is for the Sample Mean, $\bar{x} = \frac{x_1 + \dots + x_n}{n}$, with an arrow pointing from the label 'Sample Mean' to it. The middle equation is for the Sample Standard Deviation, $s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$, with an arrow pointing from the label 'Sample Standard Deviation' to it. The bottom equation is for the t-statistic, $t = \frac{\bar{x} - \mu}{s/\sqrt{n}}$, with an arrow pointing from the label 't-statistic' to it. The label 'Population Mean' is placed to the right of the t-statistic equation, with an arrow pointing from the μ term in the numerator to it. Each equation is followed by the text 'Equation by author in LaTeX.'

Sample Mean

$$\bar{x} = \frac{x_1 + \dots + x_n}{n}$$

Equation by author in LaTeX.

Sample Standard Deviation

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

Equation by author in LaTeX.

t-statistic

$$t = \frac{\bar{x} - \mu}{s/\sqrt{n}}$$

Equation by author in LaTeX.

Population Mean

Probability Density Function

The t-distribution is parameterised by only one value, the degrees of freedom, ν , and its [probability density function](#) looks like this:

$$f(t; \nu) = \frac{\Gamma\left(\frac{\nu+1}{2}\right)}{\sqrt{\nu\pi} \cdot \Gamma\left(\frac{\nu}{2}\right)} \left(1 + \frac{t^2}{\nu}\right)^{-\left(\frac{\nu+1}{2}\right)}$$

Equation by author in LaTeX.

Where:

- t is the random variable (the t -statistic).
- ν is the degrees of freedom, which is equal to $n-1$, where n is the sample size.
- $\Gamma(z)$ is the [gamma function](#), which is:

Characteristics

The mean is defined as follows for $\nu > 1$:

$$E(T) = 0$$

Equation by author in LaTeX.

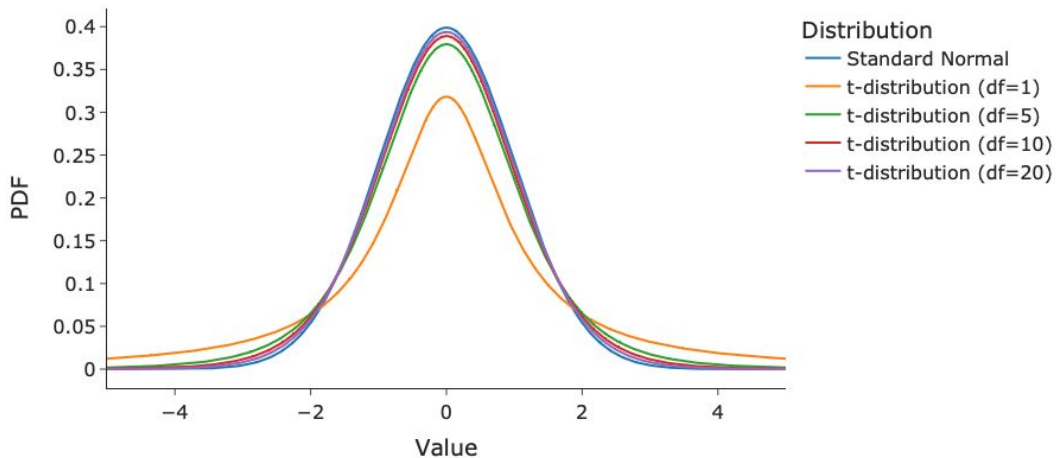
And the variance is defined as follows for $\nu > 2$:

$$Var(T) = \frac{\nu}{\nu - 2}$$

Equation by author in LaTeX.

Plots

Below is an example plot of the t-distribution as a function of various degrees of freedom and also compared to the standard normal distribution:



```
# Import packages
import numpy as np
from scipy.stats import t, norm
import plotly.graph_objects as go

# Generate data
x = np.linspace(-5, 5, 1000)
normal_pdf = norm.pdf(x, 0, 1)

# Create plot
fig = go.Figure()

# Add standard normal distribution to plot
fig.add_trace(go.Scatter(x=x, y=normal_pdf, mode='lines', name='Standard Normal'))

# Add t-distributions to plot for various degrees of freedom
for df in [1, 5, 10, 20]:
    t_pdf = t.pdf(x, df)
    fig.add_trace(go.Scatter(x=x, y=t_pdf, mode='lines', name=f't-distribution {df}'))

fig.update_layout(title='Comparison of Normal and t-distributions',
                  xaxis_title='Value',
                  yaxis_title='PDF',
                  legend_title='Distribution',
                  font=dict(size=16),
                  title_x=0.5,
                  width=900,
                  height=500,
                  template="simple_white")

fig.show()
```

Applications

- **T-test:** *The most famous application of the t -distribution is [hypothesis testing](#) through use of the t -test, which measures the statistical difference between two sample means.*
- **Confidence intervals:** *For small sample sizes (typically less than 30), it is used to compute the [confidence interval](#) for that certain statistic with increased uncertainty.*
- **Regression:** *The t -distribution is used to determine if we should add certain covariates to our regression model and calculate hypothesis tests around the significance of their coefficients.*
- **Bayesian Statistics:** *The t -distribution is sometimes used as a prior distribution in [bayesian inference](#), which can be applied in all areas of data science, particularly reinforcement learning.*

Thanks

What is the t-distribution

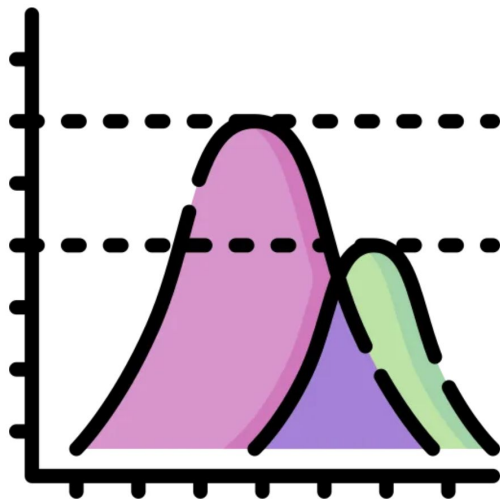
Discover the origins, theory and uses behind the famous t-distribution *



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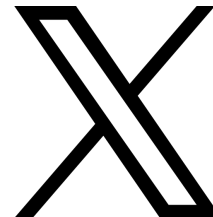
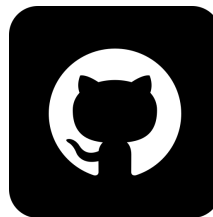
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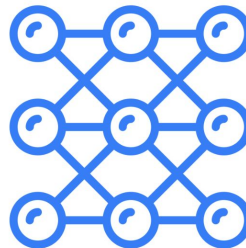


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