# F-Distribution

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#### Intuition

In this video we will cover the **F Distribution** (Fisher-Snedecor distribution) which is often used in the <u>analysis of variance (ANOVA)</u> tests and so is definitely worth knowing if you are a Data Scientist!

The F-Distribution is closely related to the Chi-Square Distribution. If you are unfamiliar with the Chi-Square Distribution, I would highly recommend checking out my previous video:



## **Origins**

A random variable is part of the F-Distribution if it satisfies the following:

$$F \sim rac{\chi_1^2/v_1}{\chi_2^2/v_2}$$

~ V

Equation produced by author in LaTeX.

$$F \sim F(v_1,v_2)$$

Equation produced by author in LaTeX.

Where  $\chi 2$  are two different Chi-Squared distributed random variables from two independent samples, which have degrees of freedom  $v_1$  and  $v_2$  respectively. In other words, it is the ratio of two Chi-Squared distributed random variables divided by their corresponding degrees of freedom.

## **Origins**

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 $F \sim F(v_1,v_2)$ 

Equation produced by author in LaTeX.

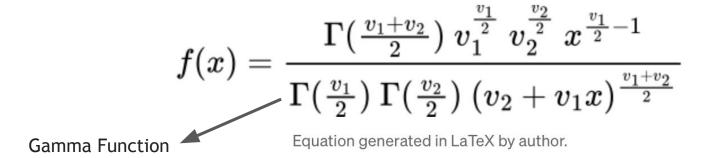
Equation produced by author in LaTeX.

Remember the degrees of freedom for the Chi-Square distribution are the number of normally distributed random variables that we square and sum up.

The distribution over all possible F values gives rise to the F-Distribution!

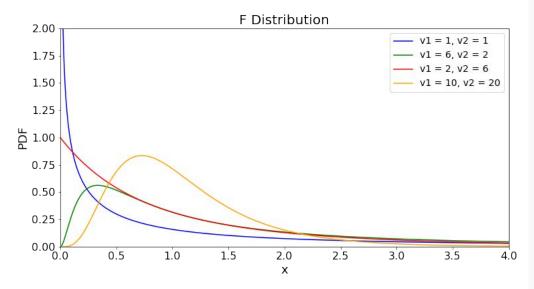
## **Probability Density Function**

The PDF for the F-distribution is this scary looking thing:



The interested reader can find a full derivation here, which includes a lot of fun things such as Jacobian matrices!

#### **Plots**



- As the degrees of freedom get larger, the distribution starts to converge towards a normal distribution.
- The distribution is often right skewed with heavy tails.
- It's always positive and is continuous.

```
# import packages
import numpy as np
from scipy.stats import f
import matplotlib.pyplot as plt
# get x values
x = np.linspace(0, 4.5, 1000)
# get F-Distributions
f1 = f(1, 1, 0)
f2 = f(6, 2, 0)
f3 = f(2, 6, 0)
f4 = f(10, 20, 0)
# plot the distributions
plt.figure(figsize=(12, 6))
plt.plot(x, f1.pdf(x), label = 'v1 = 1, v2 = 1', color = 'blue')
plt.plot(x, f2.pdf(x), label = 'v1 = 6, v2 = 2', color = 'green')
plt.plot(x, f3.pdf(x), label = 'v1 = 2, v2 = 6', color = 'red')
plt.plot(x, f4.pdf(x), label = 'v1 = 10, v2 = 20', color = 'orange')
plt.xlim(0, 4)
plt.ylim(0.0, 2)
plt.xticks(fontsize=16)
plt.yticks(fontsize=16)
plt.xlabel('x', fontsize=18)
plt.ylabel('PDF', fontsize=18)
plt.title("F Distribution", fontsize=20)
plt.legend(fontsize=14)
plt.savefig('f dist.png')
plt.show()
```

### **Thanks**

→ Member-only story

#### **F Distribution Simply Explained**

A simple and concise description of the F-Distribution



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Photo by Marin Tulard on Unsplash

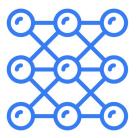
## @egorhowell







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