

# Mathematical Statistics I

## Chapter 6: Distributions Derived from the Normal Distribution

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1.  $\chi^2$  distributions

## $\chi^2$ distributions

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# Introduction

- This material comes primarily from Rice (2007, Chapter 6).
- Here, we introduce several important distributions that arise from transformations applied to normal distributions.
- Many of these distributions form the basis of traditional statistical inference procedures that are taught in introductory statistics courses.
- They are very useful in practice due to the central limit theorem: with enough observations, the limiting behavior of nearly all distributions is normal, so distributions that come from the normal distribution arise in practice as well.

- The first distribution we will consider is the  $\chi^2_1$  (Chi-square with 1 degree of freedom).

**Definition:  $\chi^2_1$  distribution**

If  $Z$  is a standard normal random variable, then  $X = Z^2$  is called the chi-square distribution with 1 degree of freedom.

- We typically use the notation  $X \sim \chi^2_1$  (in LaTeX: `\chi`).

## $\chi^2_\nu$ Distribution II

### The pdf of $\chi^2_1$

Let  $X$  follow a  $\chi^2_1$  distribution. Then, the pdf of  $X$  is given by

$$f_X(x) = \frac{1}{\sqrt{2\pi}} x^{-1/2} e^{-x/2}.$$

- In Chapter 2, we previously noted that that  $f_X(x)$  is an example of a Gamma distribution.
- Specifically, the *kernel* of the Gamma density is  $x$  raised to some power, and  $e$  raised to some multiple of  $x$ :

$$f_{\text{Gamma}}(x) \propto x^{\alpha-1} e^{-\lambda x}.$$

- Thus, ignoring the constant for a moment, if  $\alpha = 1/2$ ,  $\lambda = 1/2$ , then the pdf of  $X \sim \chi^2_1$  is just this Gamma density:

$$f_X(x) \propto x^{-1/2} e^{-x/2} = x^{\alpha-1} e^{-\lambda x}.$$

- Since both functions are proper probability density functions, they have to integrate to one, so the normalizing constant *must* be the same.

## $\chi^2_\nu$ Distribution IV

- This is also easily verified. The normalizing constant of the Gamma distribution is  $\lambda^\alpha/\Gamma(\alpha)$ .
- With our specific values of  $\lambda = \alpha = 1/2$ , and recalling that  $\Gamma(1/2) = \sqrt{\pi}$ ,

$$\frac{1}{\sqrt{2\pi}} = \frac{(1/2)^{(1/2)}}{\Gamma(1/2)} = \frac{\lambda^\alpha}{\Gamma(\alpha)}$$



## References and Acknowledgements

Rice JA (2007). *Mathematical statistics and data analysis*, volume 371. 3 edition. Thomson/Brooks/Cole Belmont, CA.

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