Lab Assignment 3 Dai Li

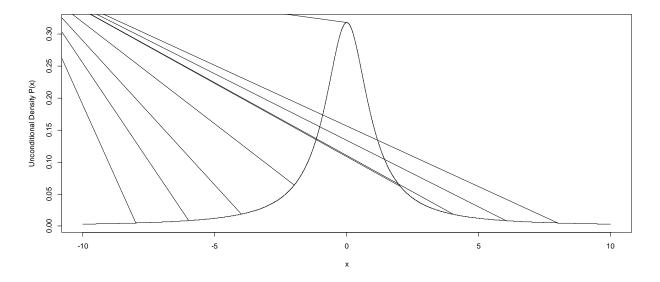
## 1. Find the Marginal Distribution p(x)

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- Normal distribution X~ N(H =2)
$f(x M, \tau) = \frac{1}{2\pi^2} (x-\mu)^2$
-Normal distribution $\chi \sim N(\mu, \sigma^2)$ $f(x \mu,\sigma) = \overline{\int_{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2}(\chi-\mu)^2}, -\infty < \chi < \infty$
Cramma distribution X~ Crammal 2. B)
$f(x \alpha,\beta) = \beta^{\alpha} \frac{1}{\Gamma(\alpha)} \chi^{\alpha-1} e^{-\beta \chi},  \chi \geqslant 0, \ \alpha,\beta > 0$
-Now, $\chi \mid \vec{t} \sim N(0, \frac{1}{2})$ , $\vec{t} \sim Gamma(\frac{2}{2}, \frac{2}{2})$
So, $\pi(\vec{z}) = (\frac{1}{2})^{\frac{1}{2}} \frac{1}{\Gamma(\frac{1}{2})} (\vec{z})^{(\frac{1}{2}-1)} e^{\frac{1}{2}\vec{z}^2}$
and, $f(x o,t) = \frac{1}{\sqrt{2n(k)}} e^{-\frac{1}{2(k)}(x-o)}$
$=\frac{\sqrt{c}}{\sqrt{z}}e^{\left(-\frac{z}{2}\chi^2\right)}$
The marginal distribution of x, pcx)
$p(x) = \int f(x \vec{t})  \chi(\vec{t})  d\vec{t}$
- Forese green Har. The marainal distribution ix Pens
- Consequently, The marginal distribution of x, Pex)
can be derived from
Can be advised from
$p(x) = \int P(x z) \pi(\pi) dz^{2}$
$=\int_{\overline{z}}^{\overline{z}} e^{\left(-\frac{z^{2}}{2}\chi^{2}\right)} \left(\frac{v}{z}\right)^{\frac{v}{2}} \frac{1}{\Gamma(\frac{v}{2})} (\overline{c})^{\left(\frac{v}{2}+1\right)} e^{-\frac{v}{2}} \overline{c}^{2} d\overline{c}$
$\frac{1}{2}$
$= \frac{(\sqrt{2})^{\frac{1}{2}}}{\int_{-1}^{1/2} \left(\frac{1}{2}\right)^{\frac{1}{2}}} \int_{0}^{\infty} \left(\frac{1}{2}\right)^{\frac{1}{2}-\frac{1}{2}} e^{-\frac{1}{2}\left(\frac{1}{2}\right)^{\frac{1}{2}} + \frac{1}{2}\nu\right)} dz^{\frac{1}{2}}$ $= \frac{(\sqrt{2})^{\frac{1}{2}}}{\int_{-1}^{1/2} \left(\frac{1}{2}\right)^{\frac{1}{2}}} \cdot \frac{\int_{-1}^{1/2} \left(\frac{1}{2}\right)^{\frac{1}{2}-\frac{1}{2}}}{\left(\frac{1}{2}\right)^{\frac{1}{2}-\frac{1}{2}}} e^{-\frac{1}{2}\left(\frac{1}{2}\right)^{\frac{1}{2}}}$
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$=\frac{(\sqrt{2})^2}{\sqrt{1+2}}\cdot\frac{1}{\sqrt{2}}$
$\left(\frac{1}{2}\right)^{2}$ $\left(\frac{1}{2}\left(x^{2}+V\right)\right)^{\frac{1}{2}}$
- P(V+1) (1-2-V+1
$=\frac{\Gamma(\frac{\vee \pm}{2})}{\sqrt{\sqrt{2}}\Gamma(\frac{\vee}{2})}\left(H^{\frac{2}{2}}\right)^{\frac{1}{2}}, \text{ which is}$
The pdf of students t distribution.

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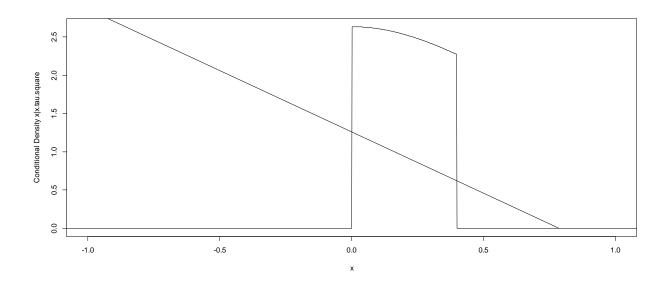
2. Let nu=1. Get a sample of 10,000 from marginal distr. of x by drawing 10,000 tau^2's and then 10,000 x's given the tau^2's. Plot sample (either histogram or density is fine). Give two names for the actual marginal distribution p(x) when nu=1. Also, compute 2.5% & 97.5% percentile points of the distribution using the random samples and compare them to the theoretical values.

The actual marginal distribution p(x) when nu=1 is plotted below:



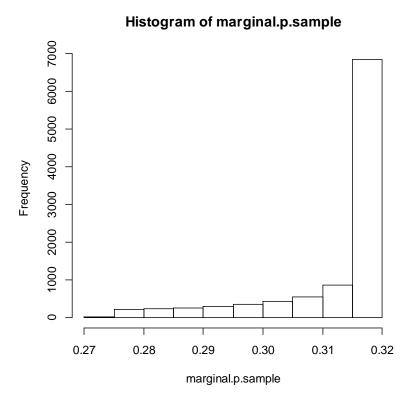
Which is a <u>Students' T Distribution with parameter Nu=1</u>. As, Nu=1, it is also a Cauchy Distribution with location parameter=0 and a scale parameter.

While using the 10000 sample points, we can draw out the theoretical conditional density of p(x|tau.square), which is plotted below:



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Which is a Student's T distribution truncated to [min(x.tau.square.sample), max(x.tau.square.sample)], whose p(x)sample value histogram is also plotted below:



The 2.5% & 97.5% percentile points of theoretical p(x) are (-12.7062, 12.7062). While, the 2.5% & 97.5% percentile points of sampled p(x) are (0.009560445, 0.3877215).

3. Use Kolmogorov-Smirnov test (ks.test in R) to test whether your observed distribution is equal to a t(df=1). Report p-value. What is the conclusion of the test?

Under null-hypothesis, p-value < 2.2e-16,

Using the sample points, D= 0.5853.

Conclusion of the test: R eject the hypothesis-the observed distribution is not equal to a t(df=1).

4. Does the Central Limit Theorem hold for the mean of a sample from p(x) when nu=1? What about nu=2? nu=3? Why or why not? A quick explanation will do; an involved proof is NOT required.

Given nu=1,2,or 3, there are finite value of X, so that there are finite value of mean E(x) and variance Var(x), and thus, the Central Limit Theorem hold for the sampled p(x).