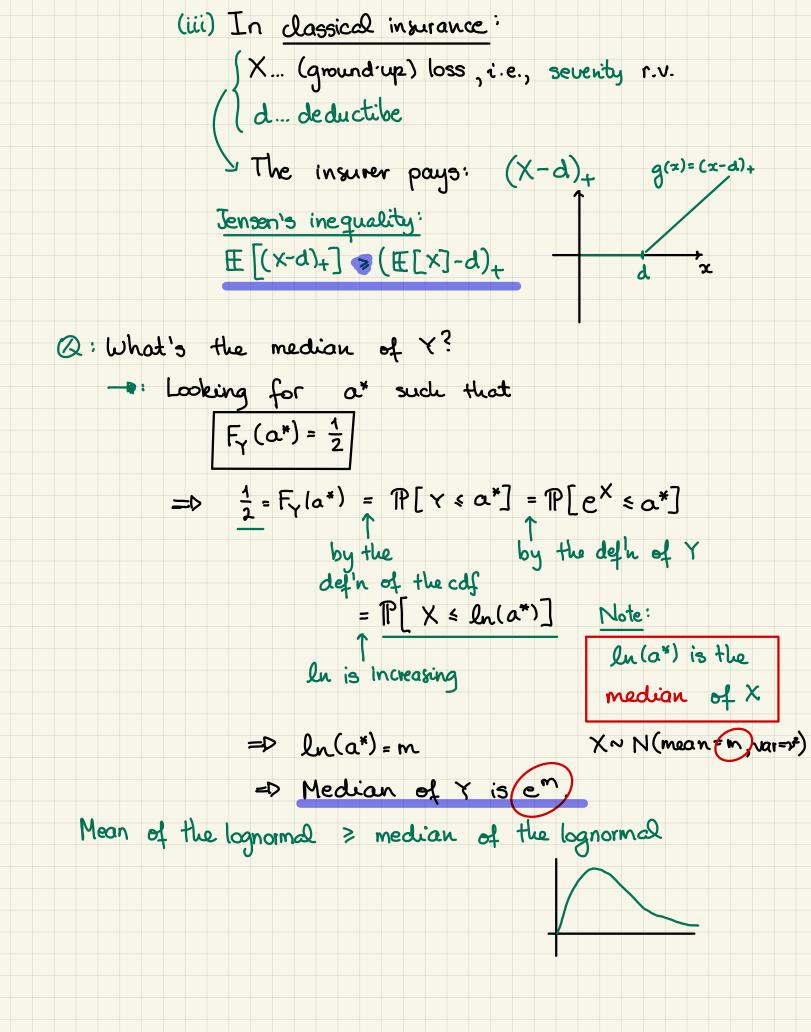
```
M339 W: March 1st, 2021.
Log Normal Distribution.
 Defn. Let X ~ Normal (mean = m, variance = 22).
           Define Y = ex.
           We say that Y is lognormally distributed.
          \mathbb{F}[Y] = \mathbb{F}[e^{X}] = \mathbb{F}[e^{m+\nu^{2}}] = e^{m} \cdot \mathbb{F}[e^{\nu^{2}}] = e^{m+\frac{\nu^{2}}{2}}
                   Let Z \sim N(0,1): X = m + v \cdot Z M_z(t) = e^{t^2/2}
                    E[X] = m
           Careat: E[ex] > e E[x]
                   This is a special case of Jensen's Inequality.
         If X is a random voviable
         and g is a convex function such that
             g(X) is well defined
             and E[g(x)] exists, then
                     \mathbb{E}[g(X)] \geqslant g(\mathbb{E}[X])
 Examples. (i) g(x) = |x|
= \emptyset \mathbb{E}[|x|] \geqslant |\mathbb{E}[x]
               (ii) Consider a European put option w/ strike K.

The expected payoff is: Payoff 1

(Vect)

(Vect)
                 E ((K-SCT))+)
                            > (K-E[S(T)])+
```



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Log-Normal Stock Prices.

S(t), t \ge 0 ... the time t stock price

R(0,t), t \ge 0 ... realized returns.

In particular, @ time T:

S(T) = S(0)e^{R(0,T)} \iff R(0,T) = ln\left(\frac{S(T)}{S(0)}\right)

We settled on the normal distribution to model realized returns, i.e.,

R(0,T) \sim Normal(mean = m, variance = 2^2).

=D S(T) is log normally distributed.
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