Лабораторная работа

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Задание 1

Описание задания 1

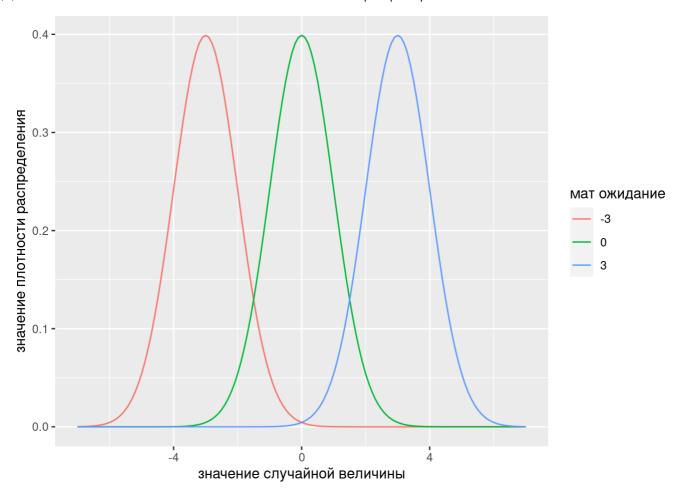
- 1. Рассмотреть 3 распределения, для каждого необходимо построить график плотности распределения (если плотность распределения неизвестна, то по характеристической функции найти плотность), функции распределения, характеристической функции.
- 2. Провести анализ параметров распределения и графиков плотности распределения (одни параметры изменяются, остальные фиксированые).
- 3. Вычислить семиинварианты и найти E(x), V(x), S(x) и K(x) через них.
- 4. Найти квантили: верхний, средний (медиана), нижний и показать как меняются квантили при изменении параметров распределения.

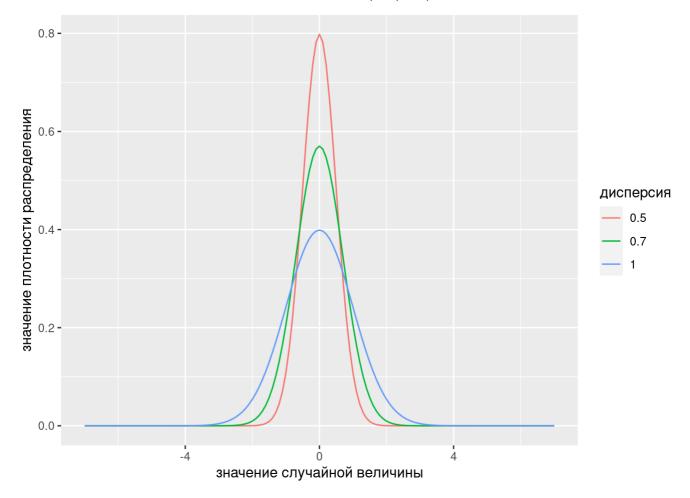
```
library(ggplot2)
library(stabledist)
library(ggforce)
library(SymTS)
library(comprehenr)
library(grid)
library(gridExtra)
library(rmutil)
library(cvar)

set.seed(1234)
x <- seq(-7, 7, 0.1)</pre>
```

Нормальное распределение

```
mean1 < -3
mean2 <- 0
mean3 <- 3
sd1 < -1
sd2 < -0.7
sd3 < -0.5
distr_by_mean <- data.frame(cond = factor(rep(c(mean1, mean2, mean3), each=length</pre>
(x))),
                     rating = c(dnorm(x, mean=mean1, sd=sd1),
                                 dnorm(x, mean=mean2, sd=sd1),
                                 dnorm(x, mean=mean3, sd=sd1)), x=x)
ggplot(distr by mean, aes(x=x, y=rating, color = cond)) +
  geom line() +
  labs(col = "мат ожидание") +
  xlab("значение случайной величины") +
  ylab("значение плотности распределения")
```





Выводы:

- 1. Изменение мат ожидания смещает график плотности распределения
- 2. Изменение дисперсии сжимает, либо вытягивает график плотности распределения

```
print_quant_norm <- function(mean, sd){
  x = c(0.25, 0.5, 0.75)
  print(paste0("мат ожидание = ", mean))
  print(paste0("дисперсия = ", sd))
  quantiles <- paste(qnorm(x, mean=mean, sd=sd), collapse = ", ")
  print(paste0("квантили = ", quantiles))
}
print_quant_norm(mean1, sd1)</pre>
```

```
## [1] "мат ожидание = -3"
## [1] "дисперсия = 1"
## [1] "квантили = -3.67448975019608, -3, -2.32551024980392"
```

```
print_quant_norm(mean2, sd1)
```

```
## [1] "мат ожидание = 0"
## [1] "дисперсия = 1"
## [1] "квантили = -0.674489750196082, 0, 0.674489750196082"
```

```
print_quant_norm(mean3, sd1)
```

```
## [1] "мат ожидание = 3"
## [1] "дисперсия = 1"
## [1] "квантили = 2.32551024980392, 3, 3.67448975019608"
```

```
print_quant_norm(mean1, sd1)
```

```
## [1] "мат ожидание = -3"
## [1] "дисперсия = 1"
## [1] "квантили = -3.67448975019608, -3, -2.32551024980392"
```

```
print_quant_norm(mean1, sd2)
```

```
## [1] "мат ожидание = -3"
## [1] "дисперсия = 0.7"
## [1] "квантили = -3.47214282513726, -3, -2.52785717486274"
```

print_quant_norm(mean1, sd3)

```
## [1] "мат ожидание = -3"
## [1] "дисперсия = 0.5"
## [1] "квантили = -3.33724487509804, -3, -2.66275512490196"
```

I.
$$\mathcal{Z} \sim \mathcal{N}(\mathcal{N}, \mathcal{G}^{2})$$

$$C_{h}(\mathcal{Z}) = i^{-n} \frac{d^{n} l_{h} (\mathcal{G}_{\pi}(t))}{dt^{n}} \Big|_{t=0}$$

$$(\int_{\mathcal{X}} (t) = e^{it\mu - \frac{\mathcal{G}^{2}t^{2}}{2}}, \quad l_{h} \mathcal{G}_{\pi}(t) = it\mu - \frac{\mathcal{G}^{2}t^{2}}{2}$$

$$C_{1}(2) = \frac{1}{i} (i\mu - \mathcal{G}^{2}t)\Big|_{t=0} = \mathcal{N}$$

$$C_{2}(2) = \frac{1}{i} (-\mathcal{G}^{2})\Big|_{t=0} = \mathcal{G}^{2}$$

$$C_{3}(2) = C_{h}(2) = 0$$

$$E(2) = C_{h}(2) = \mathcal{N}$$

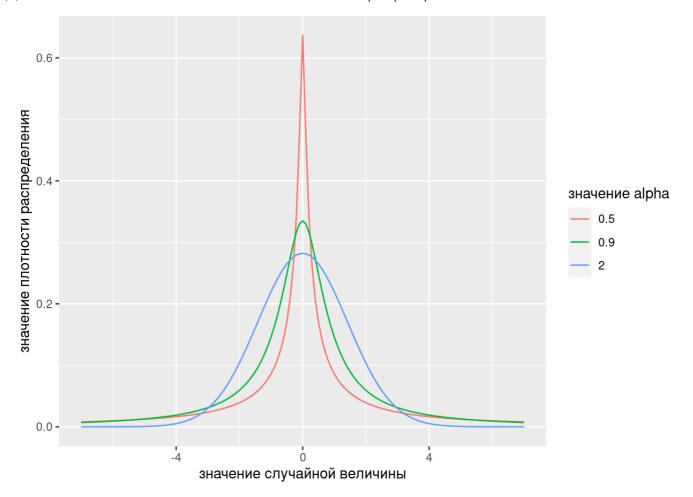
$$V(2) = C_{2}(2) = 0$$

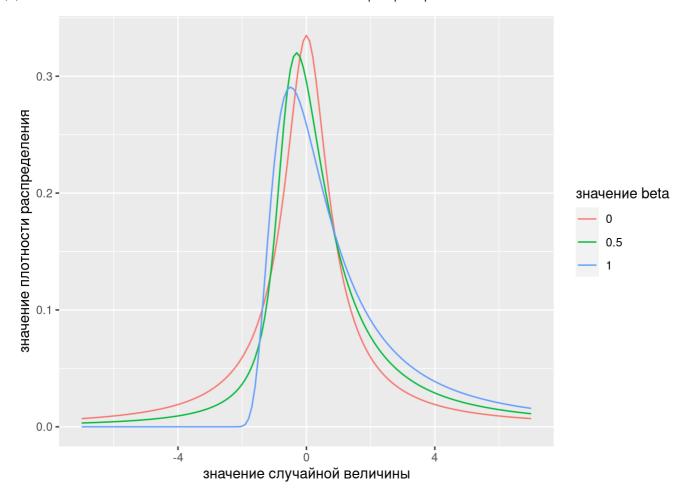
$$V(2) = C_{2}(2) = 0$$

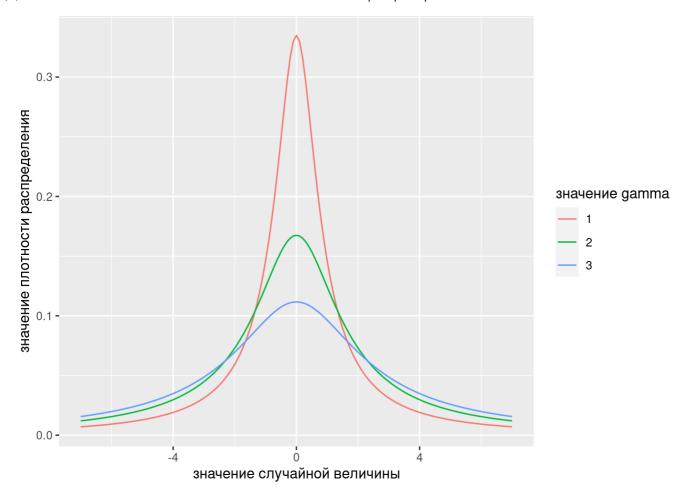
$$V(2) = C_{2}(2) = 0$$

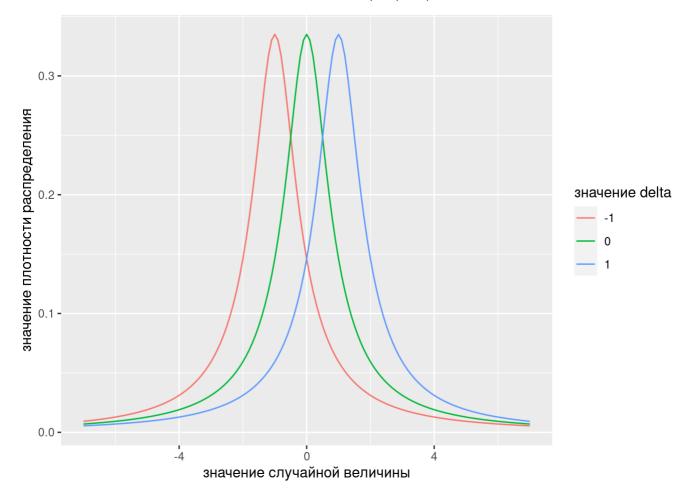
Устойчивое распределение

```
alpha1 <- 0.5
alpha2 <- 0.9
alpha3 <- 2
beta1 <- 0.5
beta2 <- 0
beta3 <- 1
qamma1 < - 2
gamma2 <- 1
gamma3 <- 3
delta1 <- -1
delta2 <- 0
delta3 <- 1
distr_by_alpha <- data.frame(cond = factor(rep(c(alpha1, alpha2, alpha3), each=length</pre>
(x))),
                             X=X,
                             rating = c(dstable(x, alpha=alpha1, beta=beta2, gamma=gam
ma2, delta = delta2),
                                        dstable(x, alpha=alpha2, beta=beta2, gamma=gam
ma2, delta = delta2),
                                        dstable(x, alpha=alpha3, beta=beta2, gamma=gam
ma2, delta = delta2)))
ggplot(distr by alpha, aes(x=x, y=rating, color = cond)) +
  geom line() +
  labs(col = "значение alpha") +
  xlab("значение случайной величины") +
  ylab("значение плотности распределения")
```









Выводы:

- 1. Параметр alpha имеет схожий эффект, что и дисперсия, но судя по формуле характеристической функции имеет более сложное влияние
- 2. Изменение параметра beta влияет на ассиметрию графика плотности распределения
- 3. Судя по графикам, параметр датта отвечает за масштаб
- 4. Параметр delta имеет схожий эффект, что и мат ожидание

```
print_quant_stable <- function(alpha, beta, gamma, delta){
    x = c(0.25, 0.5, 0.75)
    print(paste0("параметр alpha = ", alpha))
    print(paste0("параметр beta = ", beta))
    print(paste0("параметр gamma = ", gamma))
    print(paste0("параметр delta = ", delta))
    quantiles <- paste(qstable(x, alpha=alpha, beta=beta, gamma=gamma, delta = delta),
    collapse = ", ")
    print(paste0("квантили = ", quantiles))
}

print_quant_stable(alpha1, beta2, gamma2, delta2)</pre>
```

```
## [1] "параметр alpha = 0.5"

## [1] "параметр beta = 0"

## [1] "параметр gamma = 1"

## [1] "параметр delta = 0"

## [1] "квантили = -1.2838329219335, -1e-09, 1.2838329219335"
```

```
print quant stable(alpha2, beta2, gamma2, delta2)
```

```
## [1] "параметр alpha = 0.9"
## [1] "параметр beta = 0"
## [1] "параметр gamma = 1"
## [1] "параметр delta = 0"
## [1] "квантили = -1.01758385350076, -1e-09, 1.01758385350076"
```

print_quant_stable(alpha3, beta2, gamma2, delta2)

```
## [1] "параметр alpha = 2"
## [1] "параметр beta = 0"
## [1] "параметр gamma = 1"
## [1] "параметр delta = 0"
## [1] "квантили = -0.95387255240894, 0, 0.95387255240894"
```

print_quant_stable(alpha2, beta1, gamma2, delta2)

```
## [1] "параметр alpha = 0.9"
## [1] "параметр beta = 0.5"
## [1] "параметр gamma = 1"
## [1] "параметр delta = 0"
## [1] "квантили = -0.593044365717121, 0.237583627820147, 1.86295129133498"
```

print_quant_stable(alpha2, beta2, gamma2, delta2)

```
## [1] "параметр alpha = 0.9"

## [1] "параметр beta = 0"

## [1] "параметр gamma = 1"

## [1] "параметр delta = 0"

## [1] "квантили = -1.01758385350076, -1e-09, 1.01758385350076"
```

print quant stable(alpha2, beta3, gamma2, delta2)

```
## [1] "параметр alpha = 0.9"
## [1] "параметр beta = 1"
## [1] "параметр gamma = 1"
## [1] "параметр delta = 0"
## [1] "квантили = -0.380182808457033, 0.652493541760843, 2.96975428636576"
```

print quant stable(alpha2, beta2, gamma1, delta2)

```
## [1] "параметр alpha = 0.9"
## [1] "параметр beta = 0"
## [1] "параметр gamma = 2"
## [1] "параметр delta = 0"
## [1] "квантили = -2.03516770700152, -2e-09, 2.03516770700152"
```

```
print quant stable(alpha2, beta2, gamma2, delta2)
```

```
## [1] "параметр alpha = 0.9"
## [1] "параметр beta = 0"
## [1] "параметр gamma = 1"
## [1] "параметр delta = 0"
## [1] "квантили = -1.01758385350076, -1e-09, 1.01758385350076"
```

print_quant_stable(alpha2, beta2, gamma3, delta2)

```
## [1] "параметр alpha = 0.9"
## [1] "параметр beta = 0"
## [1] "параметр gamma = 3"
## [1] "параметр delta = 0"
## [1] "квантили = -3.05275156050228, -3e-09, 3.05275156050228"
```

print_quant_stable(alpha2, beta2, gamma2, delta1)

```
## [1] "параметр alpha = 0.9"
## [1] "параметр beta = 0"
## [1] "параметр gamma = 1"
## [1] "параметр delta = -1"
## [1] "квантили = -2.01758385350076, -1.000000001, 0.0175838535007597"
```

print_quant_stable(alpha2, beta2, gamma2, delta2)

```
## [1] "параметр alpha = 0.9"

## [1] "параметр beta = 0"

## [1] "параметр gamma = 1"

## [1] "параметр delta = 0"

## [1] "квантили = -1.01758385350076, -1e-09, 1.01758385350076"
```

print quant stable(alpha2, beta2, gamma2, delta3)

```
## [1] "параметр alpha = 0.9"
## [1] "параметр beta = 0"
## [1] "параметр gamma = 1"
## [1] "параметр delta = 1"
## [1] "квантили = -0.0175838535007597, 0.999999999, 2.01758385350076"
```

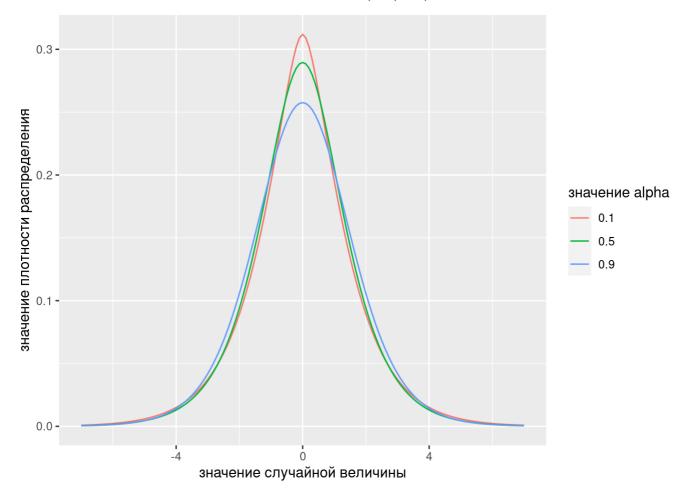
2.
$$x \sim S_{\alpha}(6, \beta, \beta)$$
, $x \neq 1, \frac{3}{2}$ (no yearline garders)

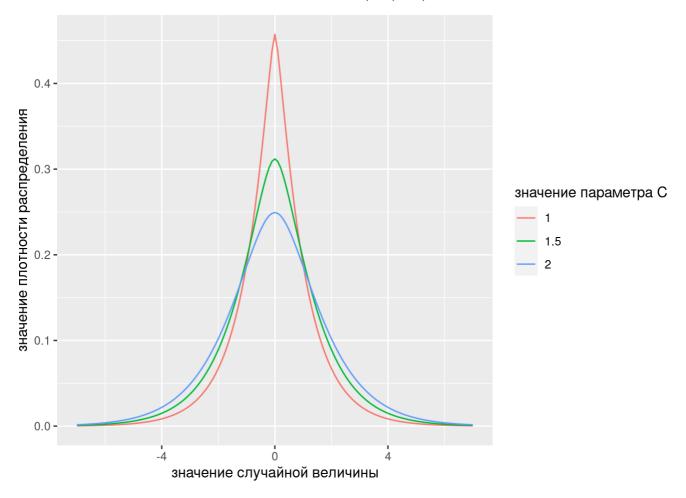
 $l_{\alpha}(f_{\alpha}(4) = i \mu t - \delta^{2d} | t |^{d} + i \delta^{d} | t |^{d} \beta t g \frac{dT}{d} = \frac{1}{2} \left[i \mu t + \delta^{2d} | t |^{d} (-1 + i \beta t g \frac{dT}{d}) \right] = \frac{1}{2} \left[i \mu t + \delta^{2d} | t |^{d} (-1 + i \beta t g \frac{dT}{d}) \right] = \mu$
 $c_{\alpha}(x) = \frac{1}{2} \left(i \mu t + \delta^{2d} | t |^{d-1} (-1 + i \beta t g \frac{dT}{d}) \right) = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right] = \frac{1}{2} \left[\frac{1}{2} \left(\frac{1}{2} + i \frac{1}{2} \frac{dT}{d} \right) \right]$

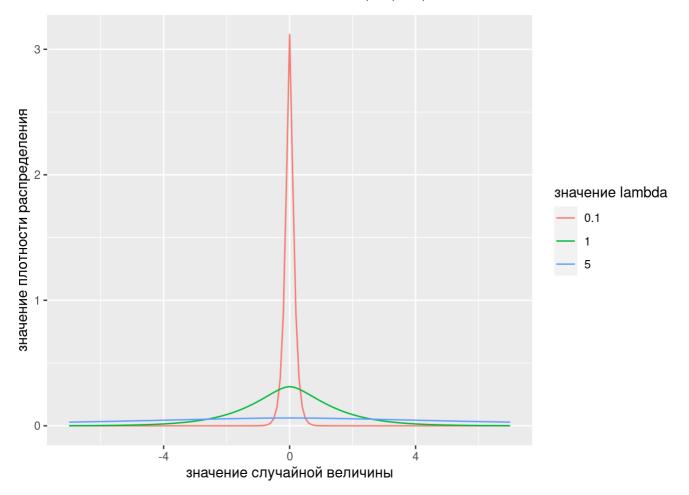
Из лекции следует, что beta параметр является параметром ассиметрами (это же следует из графиков), но по формулам выходит, что S(x) = 0

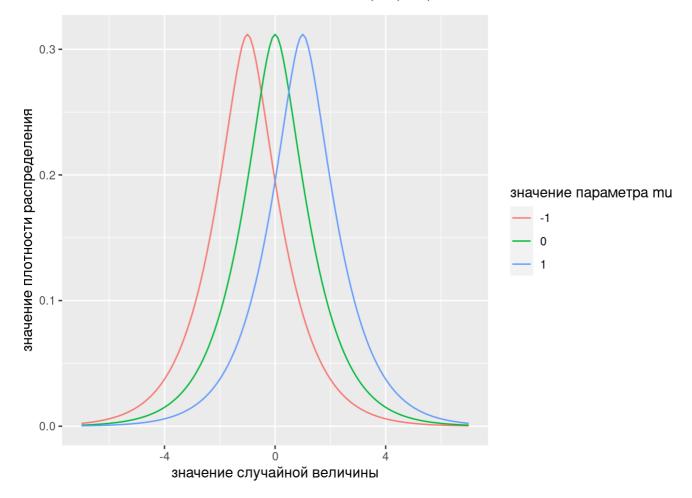
CTS распределение

```
cts alpha1 <- 0.1
cts alpha2 <- 0.5
cts alpha3 <- 0.9
lambda1 <- 0.1
lambda2 <- 1
lambda3 < -5
c1 <- 1
c2 <- 1.5
c3 <- 2
mu1 < - -1
mu2 <- 0
mu3 <- 1
cts_distr_by_cts_alpha <- data.frame(cond = factor(rep(c(cts_alpha1, cts_alpha2, cts_</pre>
alpha3), each=length(x))),
                                  X=X,
                                  rating = c(dCTS(x, alpha=cts alpha1, c=c2, ell=lambd
a2, mu = mu2),
                                         dCTS(x, alpha=cts_alpha2, c=c2, ell=lambda2,
mu = mu2),
                                         dCTS(x, alpha=cts alpha3, c=c2, ell=lambda2,
mu = mu2)))
ggplot(cts_distr_by_cts_alpha, aes(x=x, y=rating, color = cond)) +
  geom line() +
  labs(col = "значение alpha") +
  xlab("значение случайной величины") +
  ylab("значение плотности распределения")
```









Выводы:

Из графиков можно сделать вывод только о том, что все параметры кроме mu имеют схожий эффект с дисперсией. Параметр mu отвечает за сдвиг графика плотности распределения.

```
print_quant_cts <- function(cts_alpha, c, lambda, mu){
    x = c(0.25, 0.5, 0.75)
    print(paste0("параметр alpha = ", cts_alpha))
    print(paste0("параметр C = ", c))
    print(paste0("параметр lambda = ", lambda))
    print(paste0("параметр mu = ", mu))
    quantiles <- paste(qCTS(x, alpha=cts_alpha, c=c, ell=lambda, mu = mu), collapse =
", ")
    print(paste0("квантили = ", quantiles))
}
print_quant_cts(cts_alpha1, lambda2, c2, mu2)</pre>
```

```
## [1] "параметр alpha = 0.1"

## [1] "параметр C = 1"

## [1] "параметр lambda = 1.5"

## [1] "параметр mu = 0"

## [1] "квантили = -1.0441887374331, 0, 1.0441887374331"
```

```
print_quant_cts(cts_alpha2, lambda2, c2, mu2)
```

```
## [1] "параметр alpha = 0.5"
## [1] "параметр С = 1"
## [1] "параметр lambda = 1.5"
## [1] "параметр mu = 0"
## [1] "квантили = -1.10882178029597, 0, 1.10882178029597"
print quant cts(cts alpha3, lambda2, c2, mu2)
## [1] "параметр alpha = 0.9"
## [1] "параметр C = 1"
## [1] "параметр lambda = 1.5"
## [1] "параметр mu = 0"
## [1] "квантили = -1.25788637960172, 0, 1.25788637960172"
print quant cts(cts alpha1, lambda1, c2, mu2)
## [1] "параметр alpha = 0.1"
## [1] "параметр C = 0.1"
## [1] "параметр lambda = 1.5"
## [1] "параметр mu = 0"
## [1] "квантили = -0.04421845704661, 0, 0.04421845704661"
print quant cts(cts alpha1, lambda2, c2, mu2)
## [1] "параметр alpha = 0.1"
## [1] "параметр С = 1"
## [1] "параметр lambda = 1.5"
## [1] "параметр mu = 0"
## [1] "квантили = -1.0441887374331, 0, 1.0441887374331"
print quant cts(cts alpha1, lambda3, c2, mu2)
## [1] "параметр alpha = 0.1"
## [1] "параметр С = 5"
## [1] "параметр lambda = 1.5"
## [1] "параметр mu = 0"
## [1] "квантили = -2.95095887779074, 0, 2.95095887779074"
print quant cts(cts alpha1, lambda2, c1, mu2)
## [1] "параметр alpha = 0.1"
## [1] "параметр С = 1"
## [1] "параметр lambda = 1"
## [1] "параметр mu = 0"
## [1] "квантили = -0.696125824955401, 0, 0.696125824955401"
print_quant_cts(cts_alpha1, lambda2, c2, mu2)
```

```
Лабораторная работа
## [1] "параметр alpha = 0.1"
## [1] "параметр С = 1"
## [1] "параметр lambda = 1.5"
## [1] "параметр mu = 0"
## [1] "квантили = -1.0441887374331, 0, 1.0441887374331"
print quant cts(cts alpha1, lambda2, c3, mu2)
## [1] "параметр alpha = 0.1"
## [1] "параметр C = 1"
## [1] "параметр lambda = 2"
## [1] "параметр mu = 0"
## [1] "квантили = -1.3922516499108, 0, 1.3922516499108"
print quant cts(cts alpha1, lambda2, c2, mu1)
## [1] "параметр alpha = 0.1"
## [1] "параметр C = 1"
## [1] "параметр lambda = 1.5"
## [1] "параметр mu = -1"
## [1] "квантили = -2.0441887374331, -1, 0.0441887374331018"
print quant cts(cts alpha1, lambda2, c2, mu2)
## [1] "параметр alpha = 0.1"
## [1] "параметр С = 1"
## [1] "параметр lambda = 1.5"
## [1] "параметр mu = 0"
## [1] "квантили = -1.0441887374331, 0, 1.0441887374331"
print quant cts(cts alpha1, lambda2, c2, mu3)
## [1] "параметр alpha = 0.1"
## [1] "параметр С = 1"
```

```
## [1] "параметр lambda = 1.5"
## [1] "параметр mu = 1"
## [1] "квантили = -0.0441887374331018, 1, 2.0441887374331"
```

3.
$$\mathcal{X} \sim CTS$$
 prempedeneme

$$C_{1}(x) = m + C_{+} \Gamma(1-d) \lambda_{+}^{d-1} - C_{-} \Gamma(1-d) \lambda_{-}^{d-1}, n = 1$$

$$C_{1}(x) = \Gamma(n-d) \left(C_{+} \lambda_{+}^{d-1} + (-1)^{n} \left(C_{-} \lambda_{-}^{d-n} \right), n > 1$$

$$C_{1}(x) = C_{1}(x) = m + C_{+} \Gamma(1-d) \lambda_{+}^{d-1} - C_{-} \Gamma(1-d) \lambda_{-}^{d-1}$$

$$E(x) = C_{1}(x) = \Gamma(2-d) \left(C_{+} \lambda_{+}^{d-2} + (-1)^{2} C_{-} \lambda_{-}^{d-2} \right)$$

$$V(x) = C_{2}(x) = \Gamma(2-d) \left(C_{+} \lambda_{+}^{d-3} + (-1)^{3} C_{-} \lambda_{-}^{d-3} \right)$$

$$S(x) = \frac{C_{3}(x)}{C_{2}(x)^{3}/2} = \frac{\Gamma(3-d) \left(C_{+} \lambda_{+}^{d-2} + (-1)^{2} C_{-} \lambda_{-}^{d-2} \right)^{3/2}}{\left(\Gamma(2-d) \left(C_{+} \lambda_{+}^{d-2} + (-1)^{2} C_{-} \lambda_{-}^{d-2} \right)^{3/2}}$$

$$K(x) = \frac{C_{4}(x)}{C_{2}(x)^{2}} = \frac{\Gamma(4-d) \left(C_{+} \lambda_{+}^{d-4} + (-1)^{4} C_{-} \lambda_{-}^{d-2} \right)^{2}}{\left(\Gamma(2-d) \left(C_{+} \lambda_{+}^{d-4} + (-1)^{4} C_{-} \lambda_{-}^{d-2} \right)^{2}}$$

Задание 2

Описание задания 2

Исследовать зависимость меры возмущений вероятности для рассматриваемых распределений при различных 3-х функциях g(x) и различных параметрах распределения

Для вычисления мер риска будут использоваться следующие формулы:

$$\pi_g^{(1)}(X) = \int_0^\infty g(1-F_x(t)) \; dt \ \pi_g^{(2)}(X) = \int_{-\infty}^0 (g(1-F_x(t))-1) \; dt + \int_0^\infty g(1-F_x(t)) \; dt$$

```
g1 \leftarrow function(x) x^2
g2 < -function(x) (x + x ^ 6) / 2
g3 \leftarrow function(x)(x + x ^ (6 / 5)) * exp(1 - x) / 2
pil <- function(dist_func, g) {</pre>
  return(integrate(
    function(t) g(1 - dist func(t)),
    lower = 0,
    upper = +Inf,
    stop.on.error = FALSE
  )$value)
}
pi2 <- function(dist func, q) {
  return(integrate(
    function(t) g(1 - dist_func(t)) - 1,
    lower = -Inf,
    upper = 0,
    stop.on.error = FALSE
  )$value + integrate(
    function(t) g(1 - dist_func(t)),
    lower = 0,
    upper = +Inf,
    stop.on.error = FALSE
  )$value)
pi mera <- function(dist func) {</pre>
  return(data.frame(
    pi1.g1 = pi1(dist func, g1),
    pi1.g2 = pi1(dist func, g2),
    pi1.g3 = pi1(dist func, g3),
    pi2.g1 = pi2(dist func, g1),
    pi2.g2 = pi2(dist func, g2),
    pi2.g3 = pi2(dist func, g3)
  ))
}
entire mera df <- data.frame()</pre>
entire mera df <- rbind(entire mera df,
  Norm m3 1 = pi mera(function(t) pnorm(t, mean = mean1, sd = sd1)),
  Norm_0_1 = pi_mera(function(t) pnorm(t, mean = mean2, sd = sd1)),
  Norm_3_1 = pi_mera(function(t) pnorm(t, mean = mean3, sd = sd1)),
  Norm_0_1 = pi_mera(function(t) pnorm(t, mean = mean2, sd = sd1)),
  Norm 0 \ 0.7 = pi \ mera(function(t) \ pnorm(t, mean = mean2, sd = sd2)),
  Norm 0 \ 0.5 = pi \ mera(function(t) \ pnorm(t, mean = mean2, sd = sd3)),
  Stable_0.5_0_1_0 = pi_mera(function(t) pstable(t, alpha=alpha1, beta=beta2, gamma=g
amma2, delta = delta2)),
  Stable_0.9_0_1_0 = pi_mera(function(t) pstable(t, alpha=alpha2, beta=beta2, gamma=g
amma2, delta = delta2)),
  Stable 2 0 1 0 = pi mera(function(t) pstable(t, alpha=alpha3, beta=beta2, gamma=gam)
ma2, delta = delta2)),
  Stable 0.9 0.5 1 0 = pi mera(function(t) pstable(t, alpha=alpha2, beta=beta1, gamma
=gamma2, delta = delta2)),
  Stable 0.9 0 1 0 = pi mera(function(t) pstable(t, alpha=alpha2, beta=beta2, gamma=g
amma2, delta = delta2)),
  Stable_0.9_1_1_0 = pi_mera(function(t) pstable(t, alpha=alpha2, beta=beta3, gamma=g
amma2, delta = delta2)),
  Stable_0.9_0_2_0 = pi_mera(function(t) pstable(t, alpha=alpha2, beta=beta2, gamma=g
amma1, delta = delta2)),
```

```
Stable_0.9_0_1_0 = pi_mera(function(t) pstable(t, alpha=alpha2, beta=beta2, gamma=g
amma2, delta = delta2)),
Stable_0.9_0_3_0 = pi_mera(function(t) pstable(t, alpha=alpha2, beta=beta2, gamma=g
amma3, delta = delta2)),
Stable_0.9_0_1_m1 = pi_mera(function(t) pstable(t, alpha=alpha2, beta=beta2, gamma=g
gamma2, delta = delta1)),
Stable_0.9_0_1_0 = pi_mera(function(t) pstable(t, alpha=alpha2, beta=beta2, gamma=g
amma2, delta = delta2)),
Stable_0.9_0_1_1 = pi_mera(function(t) pstable(t, alpha=alpha2, beta=beta2, gamma=g
amma2, delta = delta3))
)
entire_mera_df[,1:3]
```

```
##
                            pil.gl
                                          pil.g2
                                                        pil.g3
## Norm m3 1
                      2.667987e-07 1.910772e-04 0.0006356476
## Norm 0 1
                      1.168475e-01 2.009043e-01
                                                  0.7130115382
## Norm 3 1
                      2.436574e+00 2.367732e+00
                                                  3.5957225802
                      1.168475e-01 2.009043e-01
## Norm 0 11
                                                  0.7130115382
## Norm 0 0.7
                      8.179324e-02 1.406330e-01 0.4991080822
## Norm 0 0.5
                      5.842374e-02 1.004521e-01 0.3565057691
## Stable 0.5 0 1 0 -6.707573e+01 4.202309e+03 -3.5478198286
## Stable_0.9_0_1_0
                      2.471886e-01 -1.541896e+00 9.5259038583
## Stable_2_0_1_0
                      1.652473e-01 2.841215e-01 1.0083505875
## Stable 0.9 0.5 1 0 4.629453e-01 3.726677e+00 13.4330490749
## Stable 0.9 0 1 01
                      2.471886e-01 -1.541896e+00 9.5259038583
                      7.727096e-01 5.462815e+00 19.4240461060
## Stable 0.9 1 1 0
## Stable_0.9_0_2_0
                      4.943792e-01 -3.083305e+00 19.0519031281
## Stable 0.9 0 1 02
                      2.471886e-01 -1.541896e+00 9.5259038583
## Stable 0.9 0 3 0
                      7.415671e-01 -4.625965e+00 27.6561876102
                      1.133942e-01 -1.723033e+00 8.9148830274
## Stable 0.9 0 1 ml
## Stable 0.9 0 1 03
                      2.471886e-01 -1.541896e+00 9.5259038583
## Stable 0.9 0 1 1
                      6.637966e-01 -1.180078e+00 10.3999108830
```

entire mera df[,4:6]

```
##
                             pi2.gl
                                           pi2.g2
                                                         pi2.g3
## Norm_m3_1
                      -3.564190e+00 -3.633603e+00 -2.404316e+00
## Norm 0 1
                      -5.641896e-01 -6.336032e-01
                                                  5.956842e-01
## Norm 3 1
                                    2.366397e+00
                                                  3.595684e+00
                      2.435810e+00
## Norm 0 11
                      -5.641896e-01 -6.336032e-01
                                                  5.956842e-01
## Norm_0_0.7
                      -3.949327e-01 -4.435222e-01 4.169790e-01
## Norm 0 0.5
                     -2.820948e-01 -3.168016e-01
                                                 2.978421e-01
## Stable_0.5_0_1_0
                     -3.368873e+04 -1.134722e+05 -1.903038e+04
## Stable_0.9_0_1_0
                      6.671323e+00 1.061067e+01
                                                  9.678947e+00
## Stable 2 0 1 0
                      -7.978846e-01 -8.960502e-01
                                                  8.424247e-01
## Stable_0.9_0.5_1_0 3.531908e+00 9.507754e+00
                                                  1.348990e+01
## Stable 0.9 0 1 01
                                                  9.678947e+00
                      6.671323e+00 1.061067e+01
## Stable_0.9_1_1_0
                      3.557460e-01 4.927523e+00
                                                  1.936652e+01
## Stable_0.9_0_2_0
                      1.333921e+01
                                    2.122142e+01
                                                   1.935799e+01
## Stable_0.9_0_1_02
                      6.671323e+00 1.061067e+01
                                                  9.678947e+00
## Stable 0.9 0 3 0
                      2.000881e+01
                                    3.183101e+01
                                                   2.811532e+01
## Stable 0.9 0 1 m1
                      5.669978e+00 9.611597e+00
                                                  8.678723e+00
## Stable_0.9_0_1_03
                      6.671323e+00
                                    1.061067e+01
                                                  9.678947e+00
## Stable_0.9_0_1_1
                      7.669495e+00
                                    1.161015e+01
                                                  1.067886e+01
```

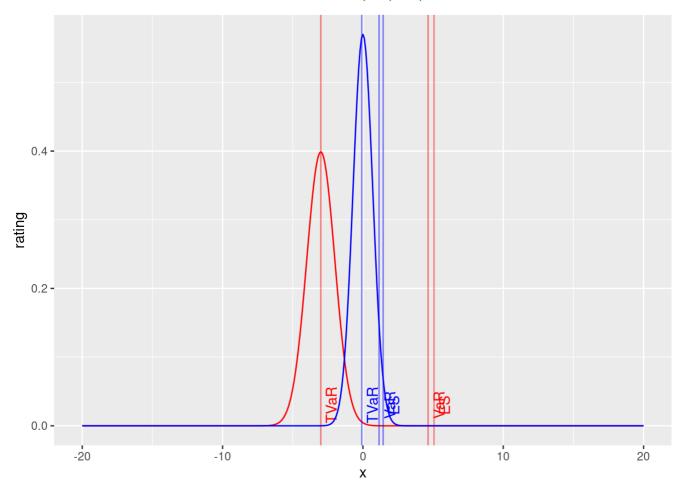
Задание 3

Описание задания 3

Проанализировать на графиках плотностей распределений расположение VaR, TVaR, ES, расположение медианы, моды и мат. ожидания

Графики для нормального распределения

```
get norm var es tvar <- function(mean, sd){</pre>
  num of samples <- 1000000
  var <- VaR(gnorm, mean=mean, sd=sd)</pre>
  es <- ES(qnorm, mean=mean, sd=sd)
  distr <- rnorm(num of samples, mean=mean, sd=sd)</pre>
  tvar <- ifelse(min(distr) < 0, mean(distr[distr < var]), mean(distr[distr > var]))
  return(c(var, es, tvar))
}
x < - seq(-20, 20, 0.1)
params1 <- get norm var es tvar(mean1, sd1)</pre>
params2 <- get norm var es tvar(mean2, sd2)</pre>
distr1 <- data.frame(rating = dnorm(x, mean=mean1, sd=sd1), x=x)</pre>
distr2 <- data.frame(rating = dnorm(x, mean=mean2, sd=sd2), x=x)</pre>
ggplot() +
  geom line(data=distr1, aes(x=x, y=rating), color = "red") +
  geom_vline(xintercept = params1[1], color="red", alpha=0.5) +
  geom text(aes(x=params1[1], label="\nVaR", y=0.03, angle=90), colour="red") +
  geom vline(xintercept = params1[2], color="red", alpha=0.5) +
  geom text(aes(x=params1[2], label="\nES", y=0.03, angle=90), colour="red") +
  geom vline(xintercept = params1[3], color="red", alpha=0.5) +
  geom text(aes(x=params1[3], label="\nTVaR", y=0.03, angle=90), colour="red") +
  geom_line(data=distr2, aes(x=x, y=rating), color = "blue") +
  geom vline(xintercept = params2[1], color="blue", alpha=0.5) +
  geom_text(aes(x=params2[1], label="\nVaR", y=0.03, angle=90), colour="blue") +
  geom vline(xintercept = params2[2], color="blue", alpha=0.5) +
  geom_text(aes(x=params2[2], label="\nES", y=0.03, angle=90), colour="blue") +
  geom vline(xintercept = params2[3], color="blue", alpha=0.5) +
  geom text(aes(x=params2[3], label="\nTVaR", y=0.03, angle=90), colour="blue")
```



Графики для устойчивого распределения

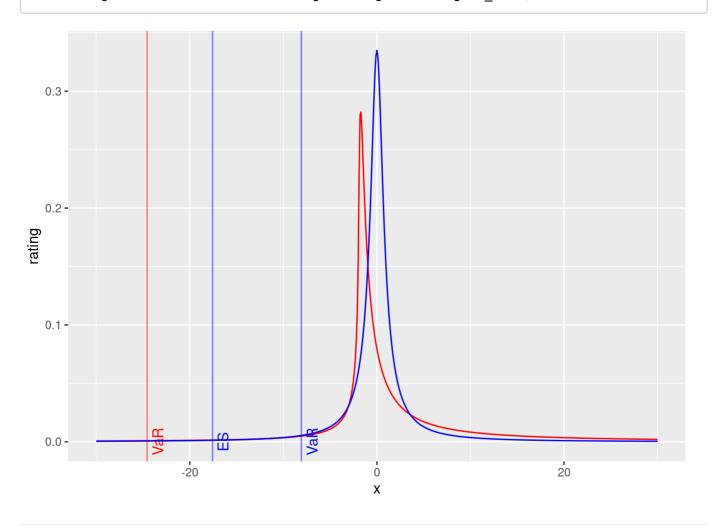
```
get stable var es tvar <- function(alpha, beta, gamma, delta){</pre>
  num of samples <- 1000000
  distr <- rstable(num of samples, alpha=alpha, beta=beta, gamma=gamma, delta=delta)
  var <- quantile(distr, 0.05)</pre>
  es <- ifelse(min(distr) < 0, mean(distr[distr < 0]), mean(distr))
  tvar <- ifelse(min(distr) < 0, mean(distr[distr < var]), mean(distr[distr > var]))
  return(c(var, es, tvar))
}
x < - seq(-30, 30, 0.1)
params1 <- get stable var es tvar(alpha=alpha1, beta=beta1, gamma=gamma1, delta=delta
params2 <- get stable var es tvar(alpha=alpha2, beta=beta2, gamma=gamma2, delta=delta
distr1 <- data.frame(rating = dstable(x, alpha=alpha1, beta=beta1, gamma=gamma1, delt</pre>
a=delta1), x=x)
distr2 <- data.frame(rating = dstable(x, alpha=alpha2, beta=beta2, gamma=gamma2, delt</pre>
a=delta2), x=x)
ggplot() +
  geom_line(data=distr1, aes(x=x, y=rating), color = "red") +
  geom vline(xintercept = params1[1], color="red", alpha=0.5) +
  geom_text(aes(x=params1[1], label="\nVaR", y=0, angle=90), colour="red") +
  geom vline(xintercept = params1[2], color="red", alpha=0.5) +
  geom_text(aes(x=params1[2], label="\nES", y=0, angle=90), colour="red") +
  geom vline(xintercept = params1[3], color="red", alpha=0.5) +
  geom text(aes(x=params1[3], label="\nTVaR", y=0, angle=90), colour="red") +
  geom line(data=distr2, aes(x=x, y=rating), color = "blue") +
  geom vline(xintercept = params2[1], color="blue", alpha=0.5) +
  geom text(aes(x=params2[1], label="\nVaR", y=0, angle=90), colour="blue") +
  geom_vline(xintercept = params2[2], color="blue", alpha=0.5) +
  geom_text(aes(x=params2[2], label="\nES", y=0, angle=90), colour="blue") +
  geom vline(xintercept = params2[3], color="blue", alpha=0.5) +
  geom text(aes(x=params2[3], label="\nTVaR", y=0, angle=90), colour="blue") +
  xlim(-30, 30)
## Warning: Removed 1 rows containing missing values (geom_vline).
## Warning: Removed 1 rows containing missing values (geom text).
## Warning: Removed 1 rows containing missing values (geom vline).
```

file:///home/max/main/univ/management_of_risk/summary.html

Warning: Removed 1 rows containing missing values (geom text).

Warning: Removed 1 rows containing missing values (geom vline).

Warning: Removed 1 rows containing missing values (geom_text).



Графики для CTS распределения

```
get cts var es tvar <- function(cts alpha, c, lambda, mu){</pre>
  num of samples <- 1000
  distr <- rCTS(num of samples, alpha=cts alpha, c=c, ell=lambda, mu = mu)
  var <- quantile(distr, 0.05)</pre>
  es <- ifelse(min(distr) < 0, mean(distr[distr < 0]), mean(distr))
  tvar <- ifelse(min(distr) < 0, mean(distr[distr < var]), mean(distr[distr > var]))
  return(c(var, es, tvar))
}
x < - seq(-30, 30, 0.1)
params1 <- get cts var es tvar(cts alpha1, c1, lambda1, mu1)</pre>
params2 <- get cts var es tvar(cts alpha2, c2, lambda2, mu2)</pre>
distr1 <- data.frame(rating = dCTS(x, alpha=cts alpha1, c=c1, ell=lambda2, mu = mu1),</pre>
distr2 <- data.frame(rating = dCTS(x, alpha=cts alpha2, c=c2, ell=lambda2, mu = mu2),</pre>
x=x)
qqplot() +
  geom line(data=distr1, aes(x=x, y=rating), color = "red") +
  geom vline(xintercept = params1[1], color="red", alpha=0.5) +
  geom_text(aes(x=params1[1], label="\nVaR", y=0, angle=90), colour="red") +
  geom vline(xintercept = params1[2], color="red", alpha=0.5) +
  geom_text(aes(x=params1[2], label="\nES", y=0, angle=90), colour="red") +
  geom vline(xintercept = params1[3], color="red", alpha=0.5) +
  geom text(aes(x=params1[3], label="\nTVaR", y=0, angle=90), colour="red") +
  geom line(data=distr2, aes(x=x, y=rating), color = "blue") +
  geom vline(xintercept = params2[1], color="blue", alpha=0.5) +
  geom text(aes(x=params2[1], label="\nVaR", y=0, angle=90), colour="blue") +
  geom_vline(xintercept = params2[2], color="blue", alpha=0.5) +
  geom text(aes(x=params2[2], label="\nES", y=0, angle=90), colour="blue") +
  geom_vline(xintercept = params2[3], color="blue", alpha=0.5) +
  geom_text(aes(x=params2[3], label="\nTVaR", y=0, angle=90), colour="blue") +
  xlim(-10, 10)
```

```
## Warning: Removed 400 row(s) containing missing values (geom_path).
## Warning: Removed 400 row(s) containing missing values (geom_path).
```

