


Number 11

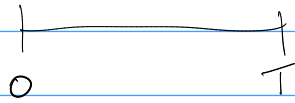
1.  сепи
маркова

$$E(\cdot), P(\cdot), V_{oe}(\cdot)$$

самый морфин у Вер-тея
и может на соб-е

2. \rightarrow TEXTURE. up learning the
processes yang ada

$$\int_0^t \lambda_u \, dW_u$$



3. ыр-ва логиче гана.

CI 95% für y_{t+2}

4. MGF ☺

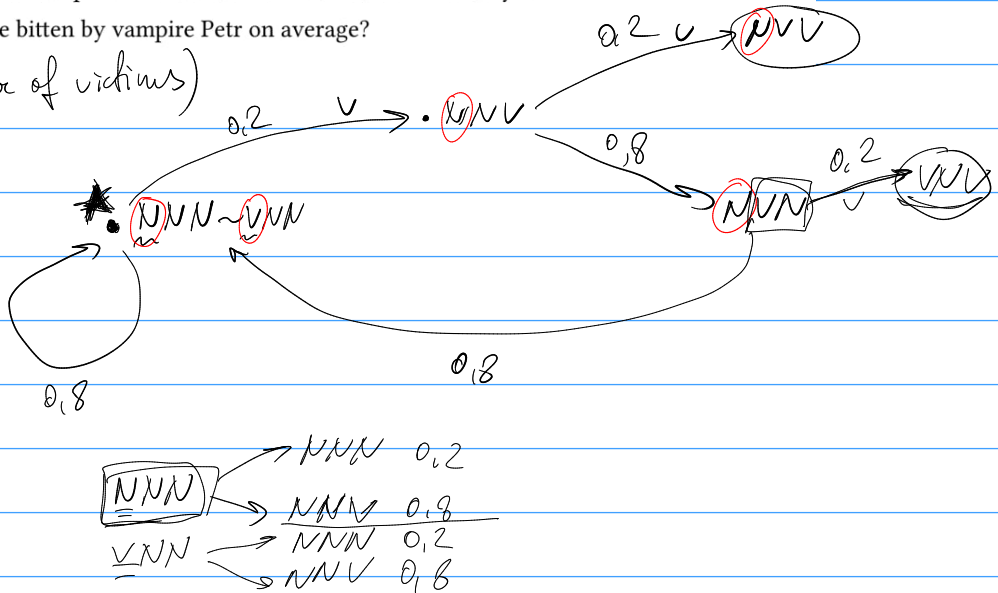
6. Vampire Petr and Markov Chains.

Vampire Petr drinks blood of a new victim every day. Unfortunately 20% of the population are vaccinated against vampires. If more than one victim of the last three victims are vaccinated then Petr will be instantaneously cured and will return to the normal life.

For simplicity let's assume that the last three victims were not vaccinated.

- What is the probability that vampire Petr will be cured in the next three days?
- How many victims will be bitten by vampire Petr on average?

(c) $\text{Var}(\text{number of victims})$



Траектории

za z x o ga; vv

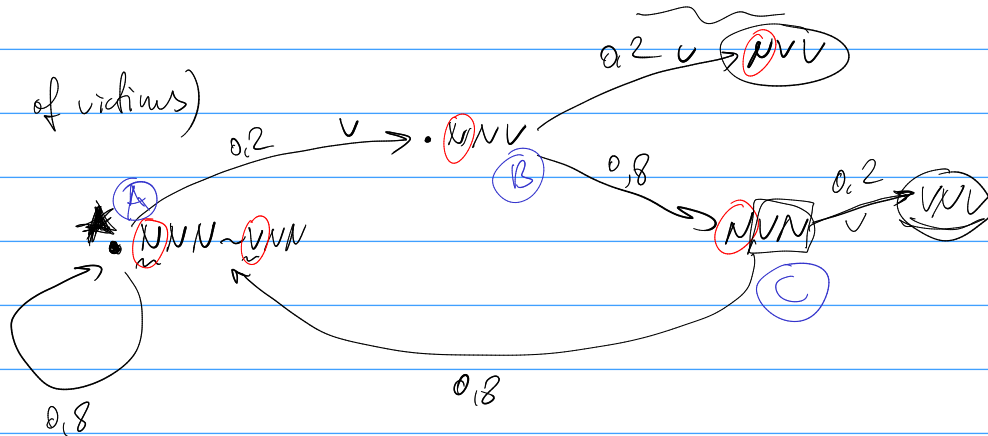
ya 3 xaga: NVV

VNV

T-спирт излучения

$$\begin{aligned} P(T \leq 3) &= P(T=2) + P(T=3) = \\ &= P(VV) + P(VVV) + P(VVVV) = \\ &= 0.2^2 + (0.2^2 \cdot 0.8) \times 2 \end{aligned}$$

N-число перов. $E(N)$



$\mu_A = E(N|A)$ - среднее ост. сч. число перов, если стоять у A.

$$\begin{cases} \mu_A = 0.8 \cdot (1 + \mu_A) + 0.2(1 + \mu_B) \\ \mu_B = 0.2 \cdot 1 + 0.8 \cdot (1 + \mu_C) \\ \mu_C = 0.2 \cdot 1 + 0.8 \cdot (1 + \mu_A) \end{cases} \Rightarrow$$

здесь законч. ф. вер. стоит

$$\begin{cases} \mu_A = 1 + 0.8\mu_A + 0.2\mu_B \\ \mu_B = 1 + 0.8\mu_C \\ \mu_C = 1 + 0.8\mu_A \end{cases} \left\{ \begin{array}{l} 0.2\mu_A = 1 + 0.2\mu_B \\ \mu_A = 5 + \mu_B \end{array} \right.$$

$$\begin{cases} \mu_B = 1 + 0.8\mu_C \\ \mu_C = 1 + 0.8(5 + \mu_B) \end{cases}$$

$$\begin{cases} \mu_B = 1 + 0.8\mu_C \\ \mu_C = 5 + 0.8\mu_B \end{cases}$$

$$\mu_C = 5 + 0.8(1 + 0.8\mu_C) \\ 5.8 + 0.64\mu_C = \mu_C$$

$$\mu_C = \frac{5.8}{0.36} = 16.11$$

$$\mu_B \approx 18.85$$

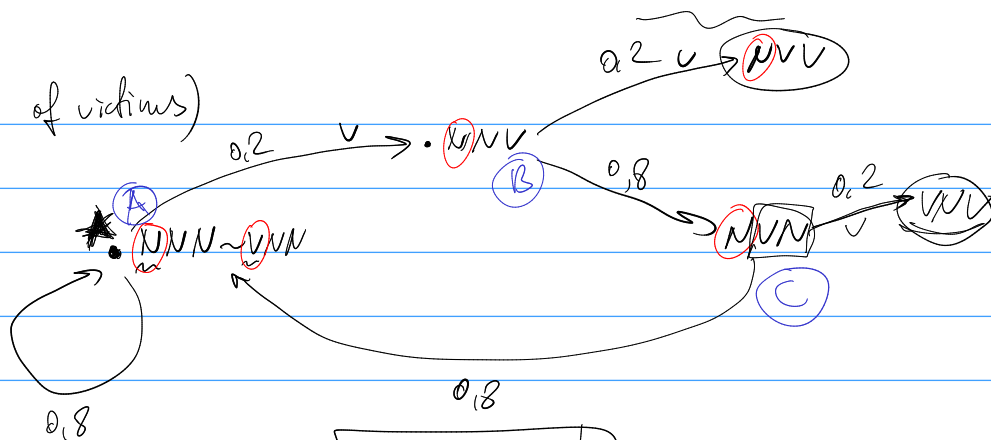
$$\mu_A \approx 23.25$$

$$\text{Var}(M) = E(M^2) - (E(M))^2$$

N-ост. сч. число перов.

$E(N|A)$, $E(N|B)$, $E(N|C)$ - экв.

$$\mu_A = E(N^2|A), \quad \mu_B = E(N^2|B), \quad \mu_C = E(N^2|C)$$



$$q_A = E(N^2 | A) = 0,2 \cdot E((N+1)^2 | B) + 0,8 \cdot E((N+1)^2 | A)$$

$$\boxed{0,2 \cdot E(N^2+1 | B) + 0,8 E(N^2+1 | A)}$$

$$(1) \quad q_A = 0,2 \left(1 + 2E(N|B) + E(N^2|B) \right) + 0,8 \left(1 + 2E(N|A) + E(N^2|A) \right)$$

$$q_A = 0,2 \left(1 + 2\underbrace{\mu_B}_{\text{здесь}} + q_B \right) + 0,8 \left(1 + 2\underbrace{\mu_A}_{\text{здесь}} + q_A \right)$$

$$q_B = 0,2 \cdot 1^2 + 0,8 \cdot E((N+1)^2 | C)$$

$$(2) \quad q_B = 0,2 + 0,8(1 + 2\mu_C + q_C)$$

$$(3) \quad q_C = 0,2 \cdot 1^2 + 0,8(1 + 2\mu_A + q_A)$$

$$\begin{cases} (1) \\ (2) \\ (3) \end{cases} \Rightarrow q_A, q_B, q_C$$

(d) Выведем функцию на MGF для N

$$MGF_N(t) = E(e^{tN})$$

$$M_A(t) = E(e^{tN} | A)$$

$$M_B(t) = E(e^{tN} | B)$$

$$M_A(t) = E(e^{tN} | A) = 0,2 \cdot E(e^{t(N+1)} | B) + 0,8 E(e^{t(N+1)} | A)$$

$$M_A(t) = 0,2 \cdot E(e^t \cdot e^{tN} | B) + 0,8 E(e^t \cdot e^{tN} | A)$$

$$M_A(t) = e^t \cdot [0,2 M_B(t) + 0,8 M_A(t)]$$

$$M_B(t) = 0.2 \cdot e^t + 0.8 E(e^{t(W+1)} | C)$$

$$M_B(t) = 0.2e^t + 0.8e^t \cdot M_C(t)$$

гип.

каждое a_t, b_t $c_t = a_t + b_t$

$a_t = 0.5a_{t-1} + u_t$ u_t - д. шум.

$b_t = 0.8b_{t-1} + w_t$ w_t - не з.б. от прошлого

u_1, u_2, \dots
 w_1, w_2, \dots

$\text{Var}\left(\begin{pmatrix} a_t \\ w_t \end{pmatrix}\right) = \begin{pmatrix} 4 & 1 \\ 1 & 9 \end{pmatrix}$

$\begin{pmatrix} u_t \\ w_t \end{pmatrix} \sim N$

u_t - д. шум.

w_t - не зависит от прошлого

u_1, u_2, \dots
 w_1, w_2, \dots

$a_{100} = 1$

$b_{100} = 2$ 95% PI для c_{102}

идея:

будущее значение = известная часть + часть, не зависящая от известных

отвечает за точечный прогноз

отвечает за ширину предельного интервала

$$\begin{aligned} c_{102} &= a_{102} + b_{102} = 0.5a_{101} + u_{102} + 0.8b_{101} + w_{102} = \\ &= 0.5(0.5a_{100} + u_{101}) + u_{102} + 0.8(0.8b_{100} + w_{101}) + w_{102} = \\ &= \boxed{0.25a_{100} + 0.64b_{100}} + (0.5u_{101} + 0.8w_{101}) + (u_{102} + w_{102}) \end{aligned}$$

$$\begin{aligned} E(c_{102} | F_{100}) &= 0.25a_{100} + 0.64b_{100} = 0.25 + 1.28 = 1.53 \\ \text{Var}(c_{102} | F_{100}) &= (0.5^2 \cdot 4 + 0.64 \cdot 9 + 2 \cdot 0.5 \cdot 0.8 \cdot 1) + \\ &\quad + (4 + 9 + 2 \cdot 1) \end{aligned}$$

95% $c_{102} \in [1.53 - 1.96 \sqrt{\text{Var}(c_{102} | F_{100})}; 1.53 + 1.96 \sqrt{\text{Var}(c_{102} | F_{100})}]$

3. (10 points) The process Y_t is defined by

$$dY_t = W_t^2 dt + W_t dW_t, \quad Y_0 = 0.$$

(a) (6 points) Find $\mathbb{E}(Y_t)$, $\mathbb{E}(Y_t W_t)$, $\mathbb{E}(Y_t W_t^2)$.

(b) (4 points) Find $\text{Var}(Y_t)$.

Идея $\int_0^t A_u dW_u$ — интеграл по виндовскому процессу W_u

(тип B_t — процесс виндовского на A_u)

$$1. \quad \mathbb{E}\left(\int_0^t A_u dW_u\right) = 0$$

Идея

$$2. \quad \text{Var}\left(\int_0^t A_u dW_u\right) = \int_0^t \mathbb{E}(A_u^2) du$$

$$\text{Зам.} \quad \text{Cov}\left(\int_0^t A_u dW_u, \int_0^t B_u dW_u\right) = \int_0^t \mathbb{E}(A_u B_u) du$$

$$3. \quad \mathbb{E}\left(\int_0^t A_u du\right) = \int_0^t \mathbb{E}(A_u) du$$

4. лемма Ито.

$$X_t = \int_0^t A_u dW_u + \int_0^t B_u du + X_0$$

нормальный процесс

$$dX_t = A_t dW_t + B_t dt$$

нормальный процесс

Лемма Ито для квадратичного процесса.

Если $Y_t = f(X_t, W_t, t)$ W_t — виндовский процесс, X_t — процесс (мартингал или предмартингал), то

dY_t можно записать так!

$$\text{то } \begin{cases} dY_t = f'_x dX_t + f'_w dW_t + f'_t dt + \\ \quad \frac{1}{2} (f''_{xx} (dX_t)^2 + f''_{ww} (dW_t)^2 + 2 f''_{xw} dX_t dW_t) \end{cases}$$

правая часть упрощается по свойствам $(dW_t)^2 = dt$

3. (10 points) The process Y_t is defined by

$$dY_t = W_t^2 dt + W_t dW_t, Y_0 = 0.$$

(a) (6 points) Find $\mathbb{E}(Y_t)$, $\mathbb{E}(Y_t W_t)$, $\mathbb{E}(Y_t W_t^2)$.

(b) (4 points) Find $\text{Var}(Y_t)$.

$$a) \quad Y_t = \int_0^t W_u^2 du + \int_0^t W_u dW_u + Y_0$$

$$\begin{aligned} \mathbb{E}(Y_t) &= \mathbb{E}\left(\int_0^t W_u^2 du + \int_0^t W_u dW_u + 0\right) = \\ &= \int_0^t \mathbb{E}(W_u^2) du + 0 + 0 = \\ &= \int_0^t u du = \boxed{\frac{t^2}{2}} \end{aligned}$$

character: W_t

$$W_t - W_s \sim N(0, t-s)$$

$$W_t \sim N(0, t)$$

$$\mathbb{E}(W_t^2) = \text{Var}(W_t) = t$$

$$\mathbb{E}(Y_t W_t) = \mathbb{E}\left[W_t \cdot \int_0^t W_u^2 du + W_t \cdot \int_0^t W_u dW_u + 0\right]$$

$$\begin{aligned} d(Y_t W_t) &= W_t \cdot dY_t + Y_t \cdot dW_t + \frac{1}{2}(2 \cdot 1 \cdot dW_t \cdot dY_t) = \\ &= W_t \cdot (W_t^2 dt + W_t dW_t) + Y_t \cdot dW_t + \underbrace{dW_t}_{\text{for}} \underbrace{\left(\frac{\partial}{\partial W} \frac{\partial}{\partial Y}\right)}_{\text{for}} \end{aligned}$$

$$d(Y_t W_t) = W_t^3 dt + W_t^2 dW_t + Y_t dW_t + W_t dt$$

$$Y_t W_t = \int_0^t W_u^3 + W_u du + \int_0^t (W_u^2 + Y_u) dW_u + Y_0 W_0$$

$$\mathbb{E}(Y_t W_t) = \int_0^t \mathbb{E}(W_u^3 + W_u) du + 0 + 0 =$$

$$dY_t dW_t = (W_t^2 dt + W_t dW_t) \cdot dW_t = 0 + W_t dt$$

$$W_u \sim N(0, u)$$

$$= 0$$

$$d(Y_t W_t^2) = f_Y' dY + f_W' dW + \frac{1}{2}(2 \cdot f_{YW} dY dW) =$$

$$d(Y_t W_t^2) = W_t^2 dY + 2W_t Y_t dW_t + 2W_t W_t dt$$

Ito's lemma

$$d(Y_t W_t^2) = W_t^2 \cdot (W_t^2 dt + W_t dW_t) + 2W_t Y_t dW_t + 2W_t^2 dt$$

$$Y_t W_t^2 = Y_0 W_0^2 + \int_0^t \dots dW_u + \int_0^t (W_u^4 + 2W_u^2) du$$

Ito's lemma

$$E(Y_t W_t^2) = \int_0^t E(W_u^4) + 2E(W_u^2) du = \int_0^t (3u^2 + 2u) du = t^3 + t^2$$

$$W_u \sim N(0, u)$$

$$E(W_u^7) = 0 \quad E(W_u^8) = u^4 \cdot E(N(0,1)^8) = u^4 \cdot 7 \cdot 5 \cdot 3 \cdot 1$$

$$E(W_u^{10}) = u^5 \cdot 9 \cdot 7 \cdot 5 \cdot 3 \cdot 1$$

$$E(W_u^2) = u \cdot 1$$

$$d) \text{Var}(Y_t) = E(Y_t^2) - (E(Y_t))^2 = E(Y_t^2) - \left(\frac{t^3}{3}\right)^2 \leftarrow \left(\frac{t^2}{2}\right)^2$$

$$d(Y_t^2) = 2Y_t \cdot dY_t + \frac{1}{2} (2 \cdot (dY_t)^2) \quad dW_t \cdot dW_t = dt$$

$$d(Y_t^2) = 2Y_t (W_t^2 dt + W_t dW_t) + (W_t^2 dt + W_t dW_t)^2 =$$

$$d(Y_t^2) = 2Y_t W_t^2 dt + \underbrace{2Y_t W_t dW_t}_{=0} + (W_t^2 dt)$$

nautilus

$$Y_t^2 = Y_0^2 + \int_0^t 2Y_u W_u^2 + W_u^2 du + \int_0^t 2Y_u W_u dW_u$$

$$E(Y_t^2) \stackrel{||}{=} \int_0^t (2 \underbrace{E(Y_u W_u^2)}_t + u) du + 0 = \int_0^t 2(u^3 + u^2) + u du$$



