

## 1. PREDSTAVAK

Problem maksimalne margeine

Linearni model bez aktivacijske funkcije

$$h(x) = \vec{w}^T \phi(\vec{x}) + w_0$$

$$y \in \{-1, +1\}$$

$$\|y - h(x)\|$$

Pretpostavka: primjeri iz  $D$  su linearno odvojivi  
(ili nemoj preklapanja u  $\mathbb{R}^n$ )

moći da je u  $x_i$  takvi da  $h(x_i) > 0$  za  $y=+1$

te  $h(x_i) < 0$  za  $y=-1$

U vrijedi  $h(x^{(i)}, y^{(i)}) \in D$ ,  $y^{(i)} h(x^{(i)}) > 0$

$\Rightarrow$  postoji  $\vec{w}$  koji zadovoljavaju  $y^{(i)} h(x^{(i)}) \geq 0$ , ali nas zanima ona max margeina

udaljenost pr. od hiperplane

$$\frac{y^{(i)} h(x^{(i)})}{\|\vec{w}\|} = \frac{y^{(i)} (\vec{w}^T \phi(x^{(i)}) + w_0)}{\|\vec{w}\|}$$

po def. margeina je jednaka udaljenosti do nebliste pr.

$$\frac{1}{\|\vec{w}\|} \min_i \{y^{(i)} (\vec{w}^T \phi(x^{(i)}) + w_0)\}$$

$\Rightarrow$  zanimaj nas  $\vec{w}$  i  $w_0$  za koje će margeina biti maksimalna

$$\operatorname{argmax}_{\vec{w}, w_0} \left\{ \frac{1}{\|\vec{w}\|} \min_i \{y^{(i)} (\vec{w}^T \phi(x^{(i)}) + w_0)\} \right\}$$

Ako je svaki primjer  $D$  linearno odvojni, onda postoji takva margeina

$\Rightarrow$  za  $x^{(i)}$  koji je nebliski margini vrijed:

$$y^{(i)} (\vec{w}^T \phi(x^{(i)}) + w_0) = 1$$

onda je  $(x^{(i)}, y^{(i)}) \in D$  mala nejednakost

$$y^{(i)} (\vec{w}^T \phi(x^{(i)}) + w_0) \geq 1, \quad i=1, \dots, N$$

zad problem se svede na  $\operatorname{argmax}_{\vec{w}, w_0} \frac{1}{\|\vec{w}\|}$  sa ograničenjem  
da je ekstremum  $\operatorname{argmin}_{\vec{w}, w_0} \frac{1}{2} \|\vec{w}\|^2$

Lagrangeova f-ja

PRIMARNI  
IZVODNI  $L(\bar{w}, w_0, \alpha) = \frac{1}{2} \|\bar{w}\|^2 - \sum_{i=1}^N \alpha_i \{y^{(i)} (\bar{w}^\top \phi(x^{(i)}) + w_0) - 1\}$

b)  $\frac{\partial L}{\partial w} = 0 \quad w = \sum_{i=1}^N \alpha_i (y^{(i)} \phi(x^{(i)}))$

$$\frac{\partial L}{\partial w_0} = 0 \quad 0 = \sum_{i=1}^N \alpha_i y^{(i)}$$

uvjetim u primarni primarni problem

$$\begin{aligned} L(\alpha) &= \frac{1}{2} \|\bar{w}\|^2 - \sum_{i=1}^N \alpha_i y^{(i)} \bar{w}^\top \phi(x^{(i)}) - \underbrace{\sum_{i=1}^N \alpha_i y^{(i)} w_0}_{0} + \sum_{i=1}^N \alpha_i \\ &= \frac{1}{2} \sum_{i=1}^N \alpha_i y^{(i)} \phi(x^{(i)})^\top \cdot \sum_{j=1}^N \alpha_j y^{(j)} \phi(x^{(j)}) + \sum_{i=1}^N \alpha_i \\ &= \sum_{i=1}^N \alpha_i - \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y^{(i)} y^{(j)} \phi(x^{(i)})^\top \phi(x^{(j)}) \end{aligned}$$

dualni problem je maksimiziran  $\sum_{i=1}^N \alpha_i - \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y^{(i)} y^{(j)} \phi(x^{(i)})^\top \phi(x^{(j)})$

ut ogranicenja na  $\alpha$   $\alpha_i \geq 0, i=1, \dots, N$

$$\sum_{i=1}^N \alpha_i y^{(i)} = 0$$

KET uvjeti:  $\alpha_i \geq 0$

$$\alpha_i (y^{(i)} h(x^{(i)}) - 1) \geq 0$$

c) kad mete granice dovoljavamo da pružaju budu na pogrešnoj strani, ali to kažnjavamo da su dulje na pogrešnoj strani.

uvode se novne varijable  $\xi_i \geq 0, i=1, \dots, N$  Vrednost je 0

$\rightarrow$  za pružanje na ispravnoj strani:  $\xi_i = 0$ , za ostale  $\xi_i = y^{(i)} h(x^{(i)})$

$\rightarrow$  pa tako će pružiti unutar marge, ali na pogrešnoj strani biti kažnjeni:  $0 \leq \xi_i < 1$

$\rightarrow$  reformulacija optimizacijskih ograničenja:

$$y^{(i)} (\vec{w}^T \phi(x^{(i)}) + w_0) \geq 1 - \xi_i, i=1, \dots, N$$

$\rightarrow$  cilj: maksimizati marginu i izniziti pr. na koef. str. granice

$$\frac{1}{2} \|\vec{w}\|^2 + C \sum_{i=1}^N \xi_i, \quad C > 0 \quad \begin{matrix} \text{kompromis u marge} \\ \text{izuzet korne} \end{matrix}$$

$$\underset{\vec{w}, w_0}{\operatorname{argmin}} \left\{ \frac{1}{2} \|\vec{w}\|^2 + C \sum_{i=1}^N \xi_i \right\}$$

$$\text{granice } L(\vec{w}, w_0, \xi, \alpha, \beta) = \frac{1}{2} \|\vec{w}\|^2 + C \sum_{i=1}^N \xi_i - \sum_{i=1}^N \alpha_i (y^{(i)} h(x^{(i)}) - 1 + \xi_i) - \sum_{i=1}^N \beta_i \xi_i$$

$$\frac{\partial L}{\partial \vec{w}} \Rightarrow \vec{w} = \sum_{i=1}^N \alpha_i y^{(i)} \phi(x^{(i)})$$

$$\frac{\partial L}{\partial w_0} \Rightarrow \sum_{i=1}^N \alpha_i y^{(i)} = 0$$

$$\frac{\partial L}{\partial \xi_i} \Rightarrow \alpha_i = C - \beta_i$$

$$\text{volna } L(\alpha) = \sum_{i=1}^N \alpha_i - \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N \alpha_i \alpha_j y^{(i)} y^{(j)} \phi(x^{(i)})^T \phi(x^{(j)})$$

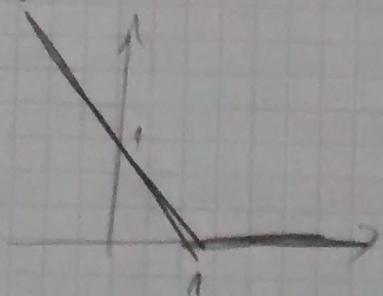
$$\text{uz ograničenja } 0 \leq \alpha_i \leq C$$

$$\sum_{i=1}^N \alpha_i y^{(i)} = 0$$

d) f-ja zglobonice

$$L(h(x), y) = \max(0, 1 - yh(x))$$

→ f-ja zglobonice je za razliku od logistike f-je jednata 0 za ispravno klasificirano primjera, i nula morgine

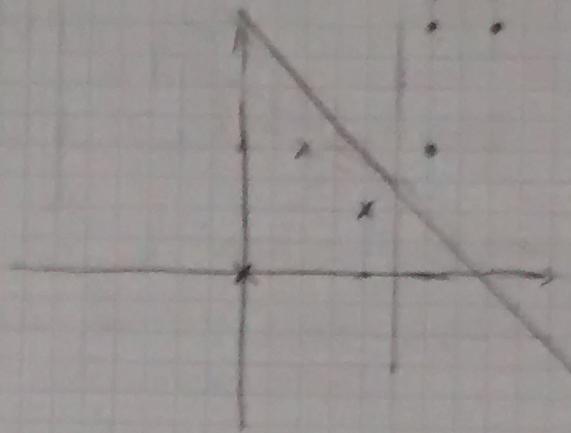


(nedostatak nije konveksna, pa nije pogodna za optimizaciju)

$$\textcircled{2} \quad D = \{(x^{(i)}, y^{(i)})\}_{i=1}^n$$

$\{(0,0), 1\}, \{(2,4), -1\}, \{(4,2), -1\}, \{(6,4), 1\}, \{(8,2), 1\}, \{(10,4), 1\}$

a)



$$h(x) = w^T \phi(x) + w_0$$

$$= x_1 + x_2 - 8$$

$$/2 \quad (\text{teg. 2x} - 6)$$

b)  $d = \frac{2}{\|w\|}$   $\rightarrow$  umkehr rückwärts

Zurück rechnen für  $x^{(1)}$  der entsprechende Vektor

c) polare Vektoren  $x^{(1)}, x^{(2)}$   $\rightarrow$  reziproker Vektor

$$w = \sum_{i=1}^n y^{(i)} (x^{(i)})$$

$$0 = \sum_{i=1}^n d_i y^{(i)}$$

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix} = d_1 (-1) \begin{bmatrix} 1 \\ 2 \end{bmatrix} + d_2 1 \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$0 = d_1 (-1) + d_2 1 \quad \Rightarrow \quad d_1 = -d_2$$

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix} = d_1 \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

$$d_1 = d_2 = \frac{1}{2}$$

$$a) \quad u_0 = \frac{1}{(2\pi)^2} \sum_{k \in \mathbb{Z}^2} (\gamma^{(0)} - \sum_{j \in \mathbb{Z}} \gamma_j) \left( P(\alpha^{(0)})^{-1} \alpha^{(0)} \right)$$

$$\begin{aligned} u_0 &= \frac{1}{4} (-1 + \left\{ \frac{1}{4}(-1)(-2)[\frac{1}{2}] + \frac{1}{4}(1)(2+1)[\frac{1}{2}] \right\}) \\ &\quad + \left( \left\{ \frac{1}{4}(-1)(-2)[1] + \frac{1}{4}(1)(2+1)[\frac{1}{2}] \right\} \right) \\ &= \frac{1}{4} \left( -\frac{1}{2} \cdot 10 + \frac{1}{2} \cdot 52 - \frac{1}{4} \cdot 20 + \frac{1}{4} \cdot 52 \right) = \\ &= \frac{1}{4} (-10 + 26 - 10 + 20) = -4 \end{aligned}$$

$$b) \quad \mathbf{x}^{(0)} = \begin{pmatrix} 5 \\ 6 \end{pmatrix}$$

Matriz representativa

$$A = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$y^{(0)} b(x^{(0)}) = 1 \cdot 20$$

$$x_1 (y^{(0)} b(x^{(0)}) - 1) = 0$$

incómodo

para res.

$$y^{(0)} \cdot (5+6-1) = 1 \cdot 20$$

$$\left\{ \begin{array}{l} y^{(0)} \geq 1 \\ y^{(0)} \neq \frac{1}{3} \end{array} \right.$$

$$y^{(0)} > \frac{1}{3}$$

$$\frac{1}{2} (y^{(0)} - 1) = 0$$

$$\frac{1}{2} y^{(0)} = \frac{1}{2}$$

$$y^{(0)} = \frac{1}{2}$$

$$y^{(0)} = \text{sgn}\left(\frac{1}{2}\right) = 1$$

$$x_1 = 1$$

1.8881, y\_2 = 0,8810,

③

- a) i model  $\Rightarrow$  prostor hipoteza kog alg. pretražuje. Hipoteza je pojednostavljeni odlike  $\Rightarrow$  DS pretražuje polpuš step hipoteze poj. model step uobičajenih stakala odlike  
 i funkcija gubitka  $\approx 0 \cdot 1$

b) optimizacijski postupak  $\Rightarrow$  pretraživanje prostora hipoteza metodom "uspon na vrh" kojem je heuristička info informacija DOBIT

(prostornost pretraživanja  $\gg$  preferiraju se kraće stakla - one koje atraktivne = većim info dobili dodaju blizu koriguju)

b) "Dobri" kriterijevi na jednom

$\gamma^{th}$  i  $f_{\text{do}}, n_r, \text{model}$

i odabir skut je razinom info dobici

$$[\text{entropija}] E(D) = \sum_{i=1}^5 p_i \log_2 p_i = -(\frac{2}{5} \log_2 \frac{2}{5} + \frac{3}{5} \log_2 \frac{3}{5} + 0.2 \log_2 0.2) \\ = +0.971$$

$$[\text{informacija}] IG(D, A) = E(D) - \sum_{\text{ve svakom } A} \frac{10,1}{10} E(D_A)$$

$$\begin{array}{lll} \text{Istra} \rightarrow 1 \text{stra} \times 4 & (\text{dan } 4) & E(\text{stra}) = 0 \\ \text{Ljubica} \times 3 & (8 \text{a} \times 3) & E(\text{ljubica}) = 0 \\ \text{Delmazina} \times 3 & (\text{dan } 2 \text{ ne } 1) & E(\text{delmazina}) = 0.92 \end{array}$$

$$IG(D, \text{ljubica}) = 0.971 - \frac{4}{10} \cdot 0 - \frac{3}{10} \cdot 0 - \frac{3}{10} \cdot 0.92 = 0.270$$

$$\begin{array}{lll} \text{II otoka} \rightarrow \text{da} \times 3 & (\text{dan } 3) & E(\text{da}) = 0 \\ \text{ne} \times 3 & (\text{dan } 2 \text{ ne } 4) & E(\text{ne}) = 0.265 \end{array}$$

$$IG(D, \text{otok}) = 0.971 - \frac{3}{10} \cdot 0 - \frac{3}{10} \cdot 0.265 = 0.2815$$

$$\begin{array}{lll} \text{III Gaydar} \rightarrow \text{privatni} \times 5 & (\text{dan } 4 \text{ ne } 1) & E(\text{privatni}) = 0.2 \\ \text{publ.} \times 5 & (\text{dan } 2 \text{ ne } 2) & E(\text{publ.}) = 0 \\ \text{zadr} \times 1 & (\text{dan } 1) & E(\text{zadr.}) = 1 \end{array}$$

$$IG(D, \text{gaydar}) = 0.971 - \frac{5}{10} \cdot 0.22 - \frac{5}{10} \cdot 1 = 0.24$$

$P(\text{velocidad})$	$\begin{matrix} 0 & 1 & 2 \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{matrix}$	$\begin{matrix} (\text{da} \times 2) \\ (\text{da} \times 3 \text{ ne} \times 3) \\ (\text{ne} \times 2) \end{matrix}$	$E(Z) = 0$
	$\begin{matrix} 0 & 1 & 2 \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{matrix}$	$\begin{matrix} (\text{da} \times 1 \text{ ne} \times 1) \\ (\text{ne} \times 1) \end{matrix}$	$E(Z) = 1$

$$P(D_1, \text{velocidad}) = 0.071 - \frac{1}{3} \cdot 1 = 0.343$$

I) Proyecto auto  $\rightarrow$  da (dañar nez)

avion  $\times 3$  (dañar nez)

autobus  $\rightarrow$  (ne - 1)

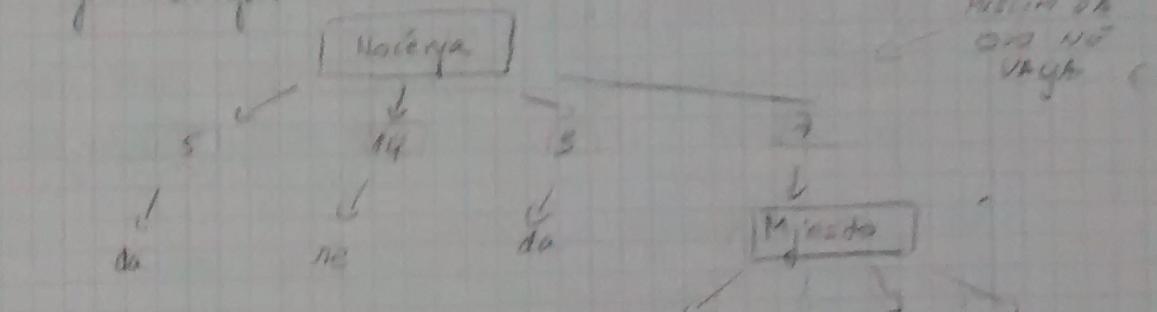
$E(\text{daño}) = 0.92$

$E(\text{dañar}) = 0.92$

$E(\text{autobus}) = 0$

$$P(D_1, \text{Proyecto}) = 0.071 - \frac{1}{3} \cdot 0.92 + \frac{2}{3} \cdot 0.02 = 0.143$$

Necesaria  $\rightarrow$  basílica



$$E(Z) = 1$$

$$P(D_1, \text{Mjeddo}) = 1 - 0.3 \cdot 0.92 - 0.276$$

$$P(D_1, \text{Dibujos}) = 1 - 0.7 \cdot 0.92 - 0.310$$

$$P(D_1, \text{Smjeddo}) = 1 - 0.5 \cdot 0.92 - 0.4 \cdot 0.276$$

$$P(D_1, \text{Proyecto}) = 1 - 0.6 \cdot 0.92 - 0.3 \cdot 0.276 = 0.172$$

$$E(13\%)$$

- c) Toda si la primera redacción es correcta  
 Segunda  $\rightarrow$  depende de la correcta de la otra y sigue el mismo patrón

Q) at 4.8

$$D = \{((x_1, 0), (1, 0, 0, 0, 0, 0), (0, 0, 0, 1, 0, 0, 0)),$$

$$(0, 0, 0, 0, 0, 0, 0), (0, 0, 0, 0, 0, 0, 0)\}$$

$$x^* = (x_1, 0) \rightarrow ? \quad x^{(1,2,3,4,5,6)} \rightarrow ?$$

Wiederholung: bei fiktiven Elementen prop. Werte zu den  
wirklichen

$$h(x_1) = \arg\max_{x \in \{0,1\}^6} \sum_{i=1}^6 \delta(x_i, y^{(i)})$$

$$\text{mit } d(x^*, y^*) = \sqrt{\sum_{i=1}^6 (x_i^* - y_i^*)^2} = \sqrt{\sum_{i=1}^6 w_i^2} = \sqrt{w_1^2 + \dots + w_6^2}$$

$$\text{Korrektur faktor } n = \frac{1}{d(x^*, y^*)^2} = \frac{n(x_1)}{\sum_{i=1}^6 w_i^2}$$

Q)  $d(0,0,0, (0,2,0)) = \sqrt{3}$

Q)  $d(0,0,0, (0,2,1)) = 1$

$d(0,0,0, (0,2,2)) = 3$

Q)  $d(0,0,0, (0,2,3)) = \sqrt{12}$

Q)  $d(0,0,0, (0,2,4)) = 5$

$d(0,0,0, (0,2,5)) = 6$

Q)  $d(0,0,0, (0,2,6)) = \sqrt{15}$

Q)  $d(0,0,0, (0,3,0)) = \sqrt{15}$

Q)  $d(0,0,0, (0,3,1)) = \sqrt{16}$

Q)  $d(0,0,0, (0,3,2)) = \sqrt{17}$

Q)  $d(0,0,0, (0,3,3)) = \sqrt{18}$

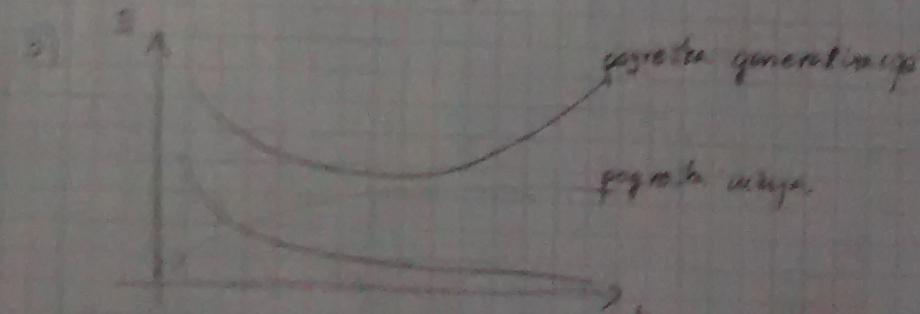
Q)  $d(0,0,0, (0,3,4)) = \sqrt{19}$

Q)  $d(0,0,0, (0,3,5)) = \sqrt{20}$

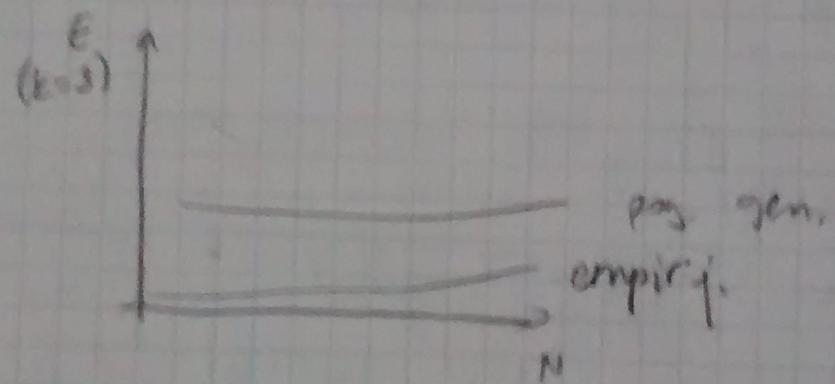
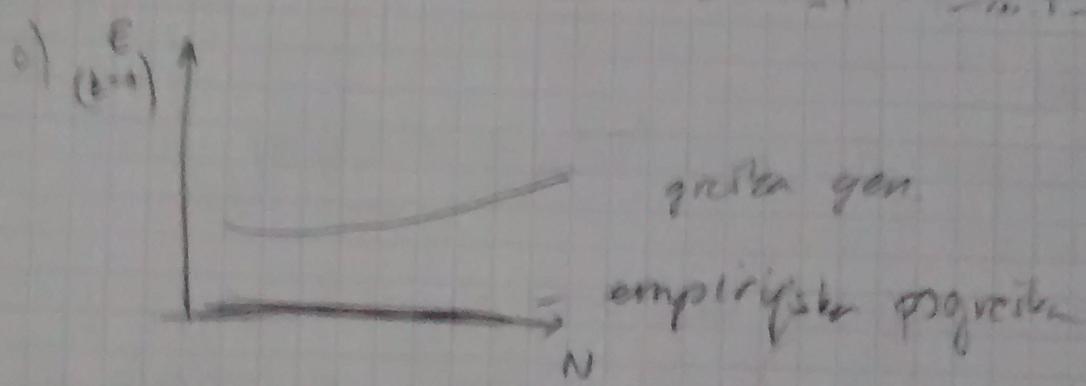
Korrektur:  $\frac{1}{3} + 1 = \frac{4}{3}$   $\rightarrow 1 \rightarrow (4,2,1)$  propada ①

Korrektur 0:  $\frac{1}{3} - \frac{1}{3} = 0$

(0,1,1) Korrekture:  $\frac{1}{3} + \frac{1}{3} = 0,0007$   
Korrektur 0:  $\frac{1}{3} - \frac{1}{3} = 0,0003 \rightarrow (0,33)$  propada ①



W-Matrix  $\beta \times \beta$  ( $d \times d$ )  $\rightarrow E(\beta) = 0$



a)

$$⑤ \text{ a) } f(y^{(i)}, h(x^{(i)})) =$$

$\{(1,1), (0,2), (2,2), (1,2), (1,1), (0,0), (1,1), (2,1), (0,1), (2,0), (2,1)\}$

$N = M$		$T_0 + F_0 + T_1 + F_1 + T_2 + F_2$ skorno		
$T_0 = 1$	$NO = 2$	0	1	2
$F_0 = 1$	$NN = 1$	1	0	1
$T_1 = 3$	$N_2 = 3$	1	3	2
$F_1 = 3$	sum negativ od redaka	pred. 1	1	1
$T_2 = 1$	sum negativ od stupnja 2		1	1
$F_2 = 2$				

preciznost = udio točno klasificiranih primjera u skupu pr.  
klasif. primjera

$$P_0 = \frac{T_0}{T_0 + F_0} = \frac{1}{2} = 0.5$$

$$P_1 = \frac{T_1}{T_1 + F_1} = \frac{3}{3+3} = \frac{1}{2}$$

$$P_2 = \frac{T_2}{T_2 + F_2} = \frac{1}{1+2} = \frac{1}{3}$$

odziv = udio točno klasificiranih primjera u skupu svih

$$R_0 = \frac{T_0}{T_0 + NO} = \frac{1}{1+2} = \frac{1}{3}$$

$$R_1 = \frac{T_1}{T_1 + NN} = \frac{3}{3+1} = \frac{3}{4}$$

$$R_2 = \frac{T_2}{T_2 + N_2} = \frac{1}{1+3} = \frac{1}{4}$$

$F_1$  mjerila = harmonička sredina preciznosti i odziva

$$F_1 = \frac{(1+\beta^2) PR}{\beta^2 P + R}$$

$$F_i = \frac{2PR_i}{P_i + R_i}$$

$$TP = \sum TP_i$$

$$FP = \sum FP_i$$

$$F_{macro} = \frac{1}{K} \sum_{i=1}^k F_i \quad F_{micro} = \frac{2TP}{P+R} = \frac{2TP}{2TP+FP+FN} \quad FN = \sum FN_i$$

$$F_0 = \frac{2 \cdot P_0 R_0}{P_0 + R_0} = \frac{2 \cdot \frac{1}{2} \cdot \frac{1}{3}}{\frac{1}{2} + \frac{1}{3}} = 0.4$$

$$F_1 = \frac{2 \cdot P_1 R_1}{P_1 + R_1} = \frac{2 \cdot \frac{1}{2} \cdot \frac{3}{4}}{\frac{1}{2} + \frac{3}{4}} = 0.6$$

$$F_2 = \frac{2 \cdot \frac{1}{3} \cdot \frac{1}{4}}{\frac{1}{3} + \frac{1}{4}} = \frac{2}{7}$$

$$F_{\text{macro}} = \frac{1}{3} \left( 0.4 + 0.6 + \frac{2}{7} \right) = \frac{3}{7} = \underline{\underline{0.423}}$$

$$TP = T0 + T1 + T2 = 5$$

$$FP = F0 + F1 + F2 = 6$$

$$FN = 6$$

$$F_1^{\text{micro}} = \frac{2.5}{2.5 + 6 + 6} = \frac{10}{22} = \frac{5}{11} = \underline{\underline{0.45}}$$

b)  $N = 1000$

ugrijedena undersampling projekta  $5 \times 5$

koliko puta provesti učenje modela?

koliko primjera u svakoj iteraciji?

1) za učenje?  $\frac{4}{5} \cdot \frac{4}{5} \cdot 1000 = \underline{\underline{640}}$

2) za projekt?  $\frac{4}{5} \cdot \frac{1}{5} \cdot 1000 = \underline{\underline{160}}$

3) za ispitivanje?  $\frac{1}{5} \cdot 1000 = \underline{\underline{200}}$