

# DOCUMENTATIE PROIECT

## Inteligența Artificială

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### Arhitectura rețelei antrenate

Pentru acest proiect, am utilizat două rețele.

Prima rețea a folosit tot setul de antrenare dat, a avut 2 straturi ascunse, unul cu 64 de perceptroni, unul cu 16 perceptroni și un strat de ieșire cu 10 perceptroni, iar funcțiile de transfer de pe aceste straturi au fost logsig pentru straturile ascunse și softmax pentru stratul de ieșire.

Cea de-a doua rețea a folosit un număr aproximativ egal de exemple din fiecare clasă a setului de antrenare (în jur de 750 exemple din fiecare clasă de la 0-9) și a păstrat arhitectura primei rețele. Pe aceasta am folosit-o exclusiv pentru îmbunătățirea outputului dat de prima rețea, astfel dacă prima rețea consideră că un exemplu face parte dintr-o clasă ce nu este 9, utilizăm răspunsul dat de această rețea echilibrată.

### Algoritmul de antrenare

Atât pentru prima rețea, cât și pentru cea de-a doua rețea am păstrat funcția de antrenare a `patternnet` și anume `trainscg`, antrenare pe baza gradientului conjugat scalat, am păstrat și preprocesările aferente pentru input și output, dar am înlocuit valorile de NaN cu media pe linie.

Pentru prima rețea am setat rata de învățare ca fiind 0.007, numărul maxim de iterații pentru care să nu existe nicio îmbunătățire pe multimea de validare l-am pus ca fiind 30 (`trainParam.max_fail`), am păstrat raportul pentru antrenare/validare/testare, iar modul în care au fost împartite exemplele a fost random. Pentru cea de-a doua rețea, cea echilibrată am setat rata de învățare ca fiind 0.008, am folosit toate exemplele pentru antrenare și am

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oprit antrenarea in momentul in care performanta a ajuns la 0.05, intrucat in urma mai multor incercari, a parut sa nu faca overfitting daca o opresc in jurul momentului acelaia.

## Performanta de clasificare medie

Am calculat performanta medie de clasificare a primei retele generand o permutare a indicilor multimii de antrenare, pe care i-am impartit in 10 submultimi disjuncte. Astfel, am obtinut urmatoarele rezultate:

Fold	Performanta
1	0.026496
2	0.026662
3	0.026744
4	0.026716
5	0.026763
6	0.026541
7	0.026461
8	0.026895
9	0.026734
10	0.02678

Asadar, performanta medie de clasificare este 0.0266792.



## Fold 2

Output Class	1	2	3	4	5	6	7	8	9	10	Accuracy
1	64 2.4%	0 0.0%	0 0.0%	0 0.0%	1 0.0%	0 0.0%	1 0.0%	0 0.0%	2 0.1%	0 0.0%	44.1% 5.9%
2	1 0.0%	29 1.1%	2 0.1%	1 0.0%	0 0.0%	0 0.0%	1 0.0%	1 0.0%	1 0.0%	5 0.2%	0.7% 9.3%
3	0 0.0%	4 0.1%	65 2.4%	1 0.0%	0 0.0%	0 0.0%	0 0.0%	5 0.2%	2 0.1%	1 0.0%	3.3% 6.7%
4	0 0.0%	2 0.1%	0 0.0%	34 1.3%	0 0.0%	1 0.0%	0 0.0%	4 0.1%	0 0.0%	14 0.5%	1.8% 8.2%
5	1 0.0%	0 0.0%	0 0.0%	0 0.0%	48 1.8%	0 0.0%	1 0.0%	0 0.0%	0 0.0%	12 0.4%	7.4% 2.6%
6	0 0.0%	0 0.0%	0 0.0%	1 0.0%	0 0.0%	44 1.6%	1 0.0%	1 0.0%	1 0.0%	1 0.0%	9.8% 0.2%
7	2 0.1%	0 0.0%	2 0.1%	0 0.0%	1 0.0%	1 0.0%	57 2.1%	0 0.0%	1 0.0%	4 0.1%	3.8% 6.2%
8	0 0.0%	2 0.1%	8 0.3%	2 0.1%	0 0.0%	1 0.0%	2 0.1%	56 2.1%	3 0.1%	3 0.1%	2.7% 7.3%
9	4 0.1%	1 0.0%	1 0.0%	0 0.0%	0 0.0%	1 0.0%	1 0.0%	3 0.1%	37 1.4%	1 0.0%	5.5% 4.5%
10	3 0.1%	19 0.7%	6 0.2%	34 1.3%	25 0.9%	21 0.8%	5 0.2%	7 0.3%	7 0.3%	199 7.4%	4.0% 6.0%
	85.3% 14.7%	0.9% 9.1%	7.4% 22.6%	6.6% 3.4%	4.0% 6.0%	3.8% 6.2%	2.6% 7.4%	2.7% 27.3%	8.5% 11.5%	8.0% 2.0%	1.0% 9.0%

### Fold 3

Validation Confusion Matrix											
Output Class	1	2	3	4	5	6	7	8	9	10	
	70	0	0	0	1	0	1	0	4	1	90.9%
	2.6%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%	0.0%	9.1%
	0	39	1	1	0	0	0	2	1	7	6.5%
	0.0%	1.5%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%	0.0%	0.3%	3.5%
	0	9	61	1	0	0	0	8	0	3	4.4%
	0.0%	0.3%	2.3%	0.0%	0.0%	0.0%	0.0%	0.3%	0.0%	0.1%	5.6%
	0	3	1	28	0	3	1	0	0	10	60.9%
	0.0%	0.1%	0.0%	1.0%	0.0%	0.1%	0.0%	0.0%	0.0%	0.4%	9.1%
	2	0	0	0	44	0	1	0	0	15	1.0%
	0.1%	0.0%	0.0%	0.0%	1.6%	0.0%	0.0%	0.0%	0.0%	0.6%	9.0%
	0	0	0	0	0	40	0	0	1	0	97.6%
	0.0%	0.0%	0.0%	0.0%	0.0%	1.5%	0.0%	0.0%	0.0%	0.0%	2.4%
	2	0	1	0	5	2	71	0	1	7	9.8%
	0.1%	0.0%	0.0%	0.0%	0.2%	0.1%	2.7%	0.0%	0.0%	0.3%	0.2%
	0	2	4	1	0	0	0	57	0	1	87.7%
	0.0%	0.1%	0.1%	0.0%	0.0%	0.0%	0.0%	2.1%	0.0%	0.0%	2.3%
	1	1	0	0	0	1	0	1	53	1	91.4%
	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	2.0%	0.0%	8.6%
	0	22	1	29	26	20	3	12	7	197	94.3%
	0.0%	0.8%	0.0%	1.1%	1.0%	0.7%	0.1%	0.4%	0.3%	4.1%	5.7%
	93.3%	1.3%	8.4%	6.7%	7.9%	0.6%	2.2%	1.2%	9.1%	7.8%	1.5%
	6.7%	48.7%	1.6%	3.3%	2.1%	9.4%	7.8%	8.7%	0.9%	2.2%	8.5%
Target Class											



## Fold 4

Validation Confusion Matrix										
Output Class	1	2	3	4	5	6	7	8	9	10
	70 2.6%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	1 0.0%	0 0.0%	3 0.1%	2 0.1%
	0 0.0%	34 1.3%	6 0.2%	2 0.1%	0 0.0%	0 0.0%	0 0.0%	2 0.1%	0 0.0%	7 0.3%
	0 0.0%	14 0.5%	63 2.4%	0 0.0%	1 0.0%	0 0.0%	1 0.0%	5 0.2%	1 0.0%	1 0.0%
	0 0.0%	1 0.0%	0 0.0%	29 1.1%	0 0.0%	1 0.0%	0 0.0%	2 0.1%	0 0.0%	12 0.4%
	1 0.0%	0 0.0%	0 0.0%	0 0.0%	44 1.6%	0 0.0%	4 0.1%	2 0.1%	1 0.0%	5 0.2%
	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	50 1.9%	1 0.0%	0 0.0%	1 0.0%	2 0.1%
	2 0.1%	1 0.0%	2 0.1%	1 0.0%	2 0.1%	0 0.0%	55 2.1%	1 0.0%	0 0.0%	4 0.1%
	0 0.0%	3 0.1%	5 0.2%	2 0.1%	1 0.0%	1 0.0%	0 0.0%	55 2.1%	3 0.1%	3 0.1%
	2 0.1%	1 0.0%	0 0.0%	0 0.0%	0 0.0%	2 0.1%	1 0.0%	1 0.0%	63 2.4%	1 0.0%
	0 0.0%	28 1.0%	6 0.2%	41 1.5%	27 1.0%	14 0.5%	2 0.1%	6 0.2%	6 0.2%	1959 73.4%
	93.3%	1.5%	6.8%	8.7%	8.7%	3.5%	4.6%	4.3%	8.8%	8.1%
	6.7%	8.5%	3.2%	1.3%	1.3%	6.5%	5.4%	5.7%	9.2%	1.9%
Target Class										

## Fold 5

Validation Confusion Matrix										
Output Class	1	2	3	4	5	6	7	8	9	10
	86	0	0	0	1	0	2	0	6	0
	3.2%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%	0.0%	0.2%	0.0%
	0	42	4	0	0	0	0	3	2	11
	0.0%	1.6%	0.1%	0.0%	0.0%	0.0%	0.0%	0.1%	0.1%	0.4%
	0	5	53	1	0	0	1	6	2	4
	0.0%	0.2%	2.0%	0.0%	0.0%	0.0%	0.0%	0.2%	0.1%	0.1%
	0	1	0	32	0	2	1	0	0	11
	0.0%	0.0%	0.0%	1.2%	0.0%	0.1%	0.0%	0.0%	0.0%	0.4%
	1	0	0	0	56	0	2	0	0	7
	0.0%	0.0%	0.0%	0.0%	2.1%	0.0%	0.1%	0.0%	0.0%	0.3%
	1	0	0	1	0	52	0	0	2	2
	0.0%	0.0%	0.0%	0.0%	0.0%	1.9%	0.0%	0.0%	0.1%	0.1%
	1	2	0	1	0	2	63	0	0	7
	0.0%	0.1%	0.0%	0.0%	0.0%	0.1%	2.4%	0.0%	0.0%	0.3%
	0	6	1	1	0	1	0	63	7	0
	0.0%	0.2%	0.0%	0.0%	0.0%	0.0%	0.0%	2.4%	0.3%	0.0%
	1	0	1	0	0	2	4	0	51	1
	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%	0.1%	0.0%	1.9%	0.0%
	1	27	2	39	19	29	5	11	4	19
	0.0%	1.0%	0.1%	1.5%	0.7%	1.1%	0.2%	0.4%	0.1%	0.7%
	94.5%	80.6%	86.9%	82.7%	83.7%	89.1%	80.8%	85.9%	88.9%	87.8%
	5.5%	19.4%	13.1%	17.3%	16.3%	10.9%	19.2%	14.1%	11.1%	12.2%
Target Class										

## Fold 6

Validation Confusion Matrix											
Output Class	1	2	3	4	5	6	7	8	9	10	
	51	0	0	0	0	0	1	0	2	0	94.4%
	1.9%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%	0.0%	5.6%
	0	30	3	2	1	0	0	2	1	7	65.2%
	0.0%	1.1%	0.1%	0.1%	0.0%	0.0%	0.0%	0.1%	0.0%	0.3%	4.8%
	0	13	66	2	0	0	0	7	0	3	72.5%
	0.0%	0.5%	2.5%	0.1%	0.0%	0.0%	0.0%	0.3%	0.0%	0.1%	7.5%
	0	2	0	39	0	2	0	1	1	8	73.6%
	0.0%	0.1%	0.0%	1.5%	0.0%	0.1%	0.0%	0.0%	0.0%	0.3%	6.4%
	1	0	0	0	48	0	1	1	0	15	72.7%
	0.0%	0.0%	0.0%	0.0%	1.8%	0.0%	0.0%	0.0%	0.0%	0.6%	7.3%
	0	1	0	1	0	41	0	1	2	2	85.4%
	0.0%	0.0%	0.0%	0.0%	0.0%	1.5%	0.0%	0.0%	0.1%	0.1%	4.6%
	1	1	0	1	2	1	71	1	0	3	87.7%
	0.0%	0.0%	0.0%	0.0%	0.1%	0.0%	2.7%	0.0%	0.0%	0.1%	2.3%
	0	3	7	1	1	0	0	53	2	1	77.9%
	0.0%	0.1%	0.3%	0.0%	0.0%	0.0%	0.0%	2.0%	0.1%	0.0%	2.1%
	2	1	0	0	0	2	0	4	62	2	84.9%
	0.1%	0.0%	0.0%	0.0%	0.0%	0.1%	0.0%	0.1%	2.3%	0.1%	5.1%
	1	24	1	35	21	25	7	7	8	196	93.8%
	0.0%	0.9%	0.0%	1.3%	0.8%	0.9%	0.3%	0.3%	0.3%	3.4%	6.2%
	91.1%	0.0%	5.7%	8.1%	5.8%	7.7%	8.8%	8.8%	9.5%	8.0%	90.7%
	8.9%	0.0%	4.3%	1.9%	4.2%	2.3%	1.3%	1.2%	0.5%	2.0%	9.3%
Target Class											



## Fold 7

Validation Confusion Matrix										
Output Class	1	2	3	4	5	6	7	8	9	10
	80 3.0%	0 0.0%	0 0.0%	0 0.0%	2 0.1%	0 0.0%	0 0.0%	0 0.0%	1 0.0%	0 0.0%
	0 0.0%	44 1.6%	2 0.1%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	9 0.3%
	0 0.0%	6 0.2%	64 2.4%	0 0.0%	0 0.0%	1 0.0%	3 0.1%	2 0.1%	1 0.0%	2 0.1%
	0 0.0%	2 0.1%	0 0.0%	25 0.9%	0 0.0%	1 0.0%	1 0.0%	1 0.0%	0 0.0%	7 0.3%
	4 0.1%	0 0.0%	0 0.0%	0 0.0%	51 1.9%	0 0.0%	4 0.1%	0 0.0%	0 0.0%	14 0.5%
	0 0.0%	0 0.0%	0 0.0%	2 0.1%	0 0.0%	38 1.4%	1 0.0%	0 0.0%	0 0.0%	0 0.0%
	0 0.0%	0 0.0%	0 0.0%	0 0.0%	3 0.1%	1 0.0%	67 2.5%	0 0.0%	0 0.0%	3 0.1%
	0 0.0%	1 0.0%	4 0.1%	1 0.0%	0 0.0%	0 0.0%	1 0.0%	69 2.6%	0 0.0%	1 0.0%
	2 0.1%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	2 0.1%	0 0.0%	0 0.0%	50 1.9%	2 0.1%
	1 0.0%	24 0.9%	3 0.1%	38 1.4%	29 1.1%	15 0.6%	5 0.2%	7 0.3%	10 0.4%	196 7.3%
Target Class										
	1	2	3	4	5	6	7	8	9	10

## Fold 8

Validation Confusion Matrix										
Output Class	1	2	3	4	5	6	7	8	9	10
	76	0	0	0	3	0	0	0	2	0
	2.8%	0.0%	0.0%	0.0%	0.1%	0.0%	0.0%	0.0%	0.1%	0.0%
	0	38	2	0	0	1	0	3	0	12
	0.0%	1.4%	0.1%	0.0%	0.0%	0.0%	0.0%	0.1%	0.0%	0.4%
	0	15	75	0	0	0	1	4	2	0
	0.0%	0.6%	2.8%	0.0%	0.0%	0.0%	0.0%	0.1%	0.1%	0.0%
	0	1	1	33	0	1	0	2	2	12
	0.0%	0.0%	0.0%	1.2%	0.0%	0.0%	0.0%	0.1%	0.1%	0.4%
	0	0	0	1	50	0	0	0	0	10
	0.0%	0.0%	0.0%	0.0%	1.9%	0.0%	0.0%	0.0%	0.0%	0.4%
	0	0	0	0	0	62	0	1	0	4
	0.0%	0.0%	0.0%	0.0%	0.0%	2.3%	0.0%	0.0%	0.0%	0.1%
	2	0	0	0	3	2	62	0	0	6
	0.1%	0.0%	0.0%	0.0%	0.1%	0.1%	2.3%	0.0%	0.0%	0.2%
	0	3	7	0	0	1	0	57	4	2
	0.0%	0.1%	0.3%	0.0%	0.0%	0.0%	0.0%	2.1%	0.1%	0.1%
	2	1	0	0	0	3	1	0	56	2
	0.1%	0.0%	0.0%	0.0%	0.0%	0.1%	0.0%	0.0%	2.1%	0.1%
	0	25	1	38	12	27	4	8	5	19
	0.0%	0.9%	0.0%	1.4%	0.4%	1.0%	0.1%	0.3%	0.2%	0.7%
	95.0%	5.8%	7.2%	5.8%	3.5%	3.9%	1.2%	6.0%	8.9%	7.6%
	5.0%	4.2%	2.8%	4.2%	6.5%	6.1%	8.8%	4.0%	1.1%	2.4%
	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%	0.0%
Target Class										

## 11

### Validation Confusion Matrix

Output Class	1	2	3	4	5	6	7	8	9	10	Accuracy
1	67 2.5%	0 0.0%	0 0.0%	0 0.0%	4 0.1%	0 0.0%	0 0.0%	0 0.0%	4 0.1%	0 0.0%	89.3%
2	0 0.0%	35 1.3%	2 0.1%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	1 0.0%	1 0.0%	9 0.3%	72.9%
3	0 0.0%	7 0.3%	61 2.3%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	11 0.4%	0 0.0%	0 0.0%	77.2%
4	0 0.0%	2 0.1%	0 0.0%	20 0.7%	0 0.0%	1 0.0%	1 0.0%	3 0.1%	0 0.0%	7 0.3%	58.8%
5	2 0.1%	0 0.0%	0 0.0%	0 0.0%	52 1.9%	0 0.0%	4 0.1%	0 0.0%	0 0.0%	8 0.3%	78.8%
6	0 0.0%	0 0.0%	0 0.0%	1 0.0%	0 0.0%	45 1.7%	1 0.0%	2 0.1%	1 0.0%	1 0.0%	88.2%
7	0 0.0%	0 0.0%	0 0.0%	0 0.0%	4 0.1%	4 0.1%	60 2.2%	2 0.1%	1 0.0%	6 0.2%	77.9%
8	0 0.0%	2 0.1%	6 0.2%	2 0.1%	0 0.0%	0 0.0%	0 0.0%	51 1.9%	3 0.1%	0 0.0%	79.7%
9	1 0.0%	1 0.0%	1 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	5 0.2%	52 1.9%	0 0.0%	86.7%
10	1 0.0%	30 1.1%	5 0.2%	30 1.1%	24 0.9%	24 0.9%	4 0.1%	9 0.3%	5 0.2%	1984 4.3%	90.9%
	5.6%	4.5%	8.7%	2.3%	8.1%	9.2%	4.3%	9.3%	2.4%	1.5%	9.1%

## Fold 10

**Validation Confusion Matrix**

Output Class	1	2	3	4	5	6	7	8	9	10	
1	93	0	0	0	2	0	1	0	3	0	93.9%
	3.5%	0.0%	0.0%	0.0%	0.1%	0.0%	0.0%	0.0%	0.1%	0.0%	5.1%
2	0	39	4	0	0	0	0	3	2	7	70.9%
	0.0%	1.5%	0.1%	0.0%	0.0%	0.0%	0.0%	0.1%	0.1%	0.3%	9.1%
3	0	12	69	1	0	0	1	2	1	0	80.2%
	0.0%	0.4%	2.6%	0.0%	0.0%	0.0%	0.0%	0.1%	0.0%	0.0%	9.8%
4	0	1	1	32	0	3	0	0	2	4	74.4%
	0.0%	0.0%	0.0%	1.2%	0.0%	0.1%	0.0%	0.0%	0.1%	0.1%	5.6%
5	0	0	0	0	51	1	2	0	0	4	87.9%
	0.0%	0.0%	0.0%	0.0%	1.9%	0.0%	0.1%	0.0%	0.0%	0.1%	2.1%
6	0	0	0	0	0	25	0	0	2	2	66.2%
	0.0%	0.0%	0.0%	0.0%	0.0%	0.9%	0.0%	0.0%	0.1%	0.1%	3.8%
7	2	0	1	1	3	0	45	1	2	4	76.3%
	0.1%	0.0%	0.0%	0.0%	0.1%	0.0%	1.7%	0.0%	0.1%	0.1%	3.7%
8	0	2	12	2	1	1	0	48	5	3	64.9%
	0.0%	0.1%	0.4%	0.1%	0.0%	0.0%	0.0%	1.8%	0.2%	0.1%	5.1%
9	1	1	0	0	0	3	1	2	68	0	89.5%
	0.0%	0.0%	0.0%	0.0%	0.0%	0.1%	0.0%	0.1%	2.5%	0.0%	0.5%
10	0	23	5	39	27	14	3	7	4	196	94.2%
	0.0%	0.9%	0.2%	1.5%	1.0%	0.5%	0.1%	0.3%	0.1%	73.7%	5.8%
	96.9%	0.0%	5.0%	2.7%	0.7%	3.2%	4.9%	6.2%	6.4%	88.8%	1.3%
	3.1%	0.0%	5.0%	7.3%	9.3%	6.8%	5.1%	23.8%	23.6%	1.2%	3.7%
Target Class	1	2	3	4	5	6	7	8	9	10	

(In tabelele matricilor de confuzie indicii sunt reprezentati de la 1 la 10 si nu de la 0 la 9 precum clasele din care fac parte exemplele.)

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În urma celor 10 matrici de confuzie am observat că un număr foarte mare de exemple care în realitate fac parte din clasa 9 sunt clasificate ca fiind în clasa 9 -- începând de la primul fold unde avem un procent de 96.9% până în ultimul fold unde ajungem la un procent de aproape 99%. Pe de altă parte, rețeaua ajunge să clasifice mult prea multe exemple ca fiind în clasa 9, astfel doar 94% din cele clasificate ca fiind în clasa 9 chiar fac parte din aceasta.

De asemenea, putem observa că cele mai multe misclasificări ale rețelei sunt exemplele din clasa 1, 3, 4 și 5 pe care le consideră ca aparțin clasei 9. Din acest motiv, am adăugat cea de-a doua rețea care conține un număr aproximativ egal de exemple din fiecare clasă.

## **Alte variante ale rețelei:**

Pentru a îmbunătăți rețeaua am mai încercat o serie de idei precum adăugarea regularizării și a normalizării standard, inițializarea ponderilor de pe fiecare strat cu valori repartizate uniform în intervalul  $[-3 * n^{-0.5}, 3 * n^{-0.5}]$ , unde  $n$  reprezintă numărul de intrări și 5-fold cross-validation pentru rețeaua cu input echilibrat.