# Globalno usavršavanje slučajne šume



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## 1. Uvod u temu

- → Globalno usavršavanje
  Usavršavanje vektora u listovima
- Globalno obrezivianje Uklanjanje čvorova u stablima radi manjeg zauzeća memorije i bolje generalizacije



### Refined-A vs Refined-E

- Maksimalna točnost modela
- Minimalno zauzeće memorije
- Razlika iteraciji zaustavljanja



# Rezultati

Dataset	Performance (Error)			Compression Ratio			
Dataset	Error Scale	RF	ADF/ARF	refined-A	refined-E	refined-A	refined-E
(c) letter	$10^{-2}$	$4.50\pm0.13$	$3.76\pm0.14$	$2.98 \pm 0.15$	$4.33\pm0.08$	2.33	30.32
(c) usps	$10^{-2}$	$6.21 \pm 0.21$	$5.60 \pm 0.16$	$5.10 \pm 0.10$	$5.69 \pm 0.15$	2.86	15.14
(c) Char74k	$10^{-2}$	$18.3 \pm 0.15$	$16.9 \pm 0.16$	$15.4 \pm 0.10$	$18.0 \pm 0.09$	1.70	37.04
(c) MNIST	$10^{-2}$	$3.14\pm0.04$	$2.73 \pm 0.05$	$2.05 \pm 0.02$	$2.95\pm0.03$	6.29	76.92
(c) covtype	$10^{-2}$	$16.4 \pm 0.10$	$15.3 \pm 0.11$	$4.11 \pm 0.04$	$15.6 \pm 0.08$	1.68	166.67
(r) abalone		$2.11\pm0.05$	$2.10 \pm 0.03$	$2.10 \pm 0.01$	$2.11\pm0.03$	12.65	16.67
(r) ailerons	$10^{-4}$	$2.01\pm0.01$	$1.98 \pm 0.01$	$1.75\pm0.02$	$1.95\pm0.02$	33.13	124.82
(r) cpusmall		$3.15 \pm 0.05$	$2.95 \pm 0.04$	$2.90 \pm 0.05$	$3.02\pm0.03$	22.73	66.53
(r) cadata	$10^{4}$	$5.50\pm0.05$	$5.40 \pm 0.05$	$5.05 \pm 0.06$	$5.36 \pm 0.05$	36.14	62.50
(r) deltaelevators	$10^{-3}$	$1.46 \pm 0.04$	$1.46 \pm 0.02$	$1.46 \pm 0.03$	$1.46 \pm 0.03$	37.04	37.04

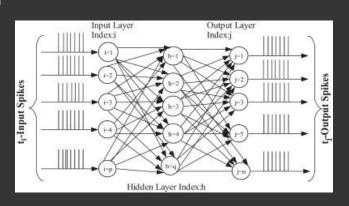
# Usporedba rezultata

Rezultati rada	Refined-A error	Refined-E error	Compression Refined-E
MNIST	0.0205	0.0295	76.92
Letter	0.0298	0.0433	30.32
Cpusmall	2.9	3.02	66.53
Abalone	2.1	2.11	16.67

Naši rezultati	Refined-A error	Refined-E error	Compression Refined-E
MNIST	0.021	0.0299	47.14
Letter	0.0328	0.0432	21.45
Cpusmall	2.89	3.138	46.12
Abalone	2.22	2.26	10.13

# Spiking neural networks Tema za usporedbu

- Neuronske mreže na koje utječe duljina signala, a ne samo snaga
- Modeliranje vremenski ovisnih signala



### Usporedba rezultata

Dataset	Performance (Error)			Compress	Compression Ratio		
Dataset	Error Scale	RF	ADF/ARF	refined-A	refined-E	refined-A	refined-E
(c) letter	$10^{-2}$	$4.50\pm0.13$	$3.76\pm0.14$	$2.98 \pm 0.15$	$4.33\pm0.08$	2.33	30.32
(c) usps	$10^{-2}$	$6.21\pm0.21$	$5.60 \pm 0.16$	$5.10 \pm 0.10$	$5.69\pm0.15$	2.86	15.14
(a) Char74k	10-2	18 3±0 15	16.0±0.16	15.4±0.10	18.040.00	1.70	37.04
(c) MNIST	$10^{-2}$	$3.14 \pm 0.04$	$2.73 \pm 0.05$	$2.05 \pm 0.02$	$2.95 \pm 0.03$	6.29	76.92
(c) cortype	10	10.4±0.10	15.3±0.11	7.111	15.0±0.00	1.00	100.07
(r) abalone		$2.11\pm0.05$	$2.10 \pm 0.03$	$2.10 \pm 0.01$	$2.11\pm0.03$	12.65	16.67
(r) ailerons	$10^{-4}$	$2.01\pm0.01$	$1.98 \pm 0.01$	$1.75\pm0.02$	$1.95\pm0.02$	33.13	124.82
(r) cpusmall		$3.15\pm0.05$	$2.95 \pm 0.04$	$2.90 \pm 0.05$	$3.02\pm0.03$	22.73	66.53
(r) cadata	$10^{4}$	$5.50\pm0.05$	$5.40 \pm 0.05$	$5.05 \pm 0.06$	$5.36\pm0.05$	36.14	62.50
(r) deltaelevators	$10^{-3}$	$1.46 \pm 0.04$	$1.46 \pm 0.02$	$1.46 \pm 0.03$	$1.46 \pm 0.03$	37.04	37.04

#### Klasifikacija EMINIST seta

#### Refined-A

Accuracy = 89.61% (8961/10000) (classification)

iteracija: 11, error: 0.1039, len(W): 130826

Minimalni error: 0.10159999999999991

#### Refined-E

Accuracy = 88.75% (8875/10000) (classification)

iteracija: 15, error: 0.1125, size: 0.2288

Minimalni error: 0.10040000000000004

Data Sets	Contenders	Accuracy (%)	Setting	Control Rate $(\gamma)$	Epochs
	Deep SNN (O'Connor and Welling, 2016)	97.80	28×28-300-300-10 🌲	-	50
	Deep SNN-BP (Lee et al., 2016)	98.71	28×28-800-10	1.5	200
	SNN-EP ♡	97.63	$28 \times 28 - 500 - 10$	1.5	25
	HM2-BP (Jin et al., 2018)	$98.84 \pm 0.02$	$28 \times 28 - 800 - 10$	-	100
MNIST	SNN-L (Rezaabad and Vishwanath, 2020)	$98.23 \pm 0.07$	28×28-1000-R28-10	12	2
	SLAYER (Shrestha and Orchard, 2018)	$98.39 \pm 0.04$	28×28-500-500-10	-	50
	SLAYER-U₁ ♣	$98.53 \pm 0.03$	28×28-500-500-10	-	-
	$SLAYER-U_2$	$98.59 \pm 0.01$	28×28-500-500-10	(-)	-
	BSNN (this work)	$99.02 \pm 0.04$	28×28-500-500-10	-0.21	50
	SKIM (Cohen et al., 2016)	92.87	2*28×28-10000-10		-
	Deep SNN-BP	98.78	2*28×28-800-10	-	200
	HM2-BP	$98.84 \pm 0.02$	2*28×28-800-10	15	60
N-MNIST	SLAYER	$98.89 \pm 0.06$	2*28×28-500-500-10	-	50
	SLAYER- $U_1$	$99.01 \pm 0.01$	2*28×28-500-500-10	-	_
	$SLAYER-U_2$	$99.07 \pm 0.02$	2*28×28-500-500-10	-	-
	BSNN (this work)	$99.24 \pm 0.12$	2*28×28-500-500-10	-0.49	50
	HM2-BP	88.99	28×28-400-400-10		15
	SLAYER	$88.61 \pm 0.17$	28×28-500-500-10	-	50
Fashion-MNIST	$SLAYER-U_1$	$90.53 \pm 0.04$	28×28-500-500-10	151	5
Fashion-MINIST	SLAYER- $U_2$	$90.61 \pm 0.02$	28×28-500-500-10	1.0	-
	ST-RSBP (Zhang and Li, 2019)	$90.00 \pm 0.13$	28×28-400-R400-10 ♦	-	30
	BSNN (this work)	$91.22 \pm 0.06$	28×28-500-500-10	-0.32	50
	eRBP (Neftci et al., 2017)	78.17	28×28-200-200-47	-	30
	HM2-BP	$84.43 \pm 0.10$	28×28-400-400-10	120	20
EMNIST	SNN-L	$83.75 \pm 0.15$	28×28-1000-R28-10	-	-
EMINIST	SLAYER	$85.73 \pm 0.16$	28×28-500-500-47	(-)	50
	SLAYER- $U_2$	$86.62 \pm 0.03$	28×28-500-500-47	-	50
	BSNN (this work)	$87.51 \pm 0.23$	28×28-500-500-47	-0.37	50

# Usporedba rezultata

Rezultati rada	Accuracy
MNIST	99.02%
EMNIST	87.51%
Fashion-MNIST	91.22%

Naši rezultati	Refined-A accuracy	Refined-E accuracy
MNIST	97.9%	97.11%
EMNIST	83.06%	82.46%
Fashion-MNIST	89.61%	88.75%



## Zaključak

- → Bolji rezultati neuronske mreže Za očekivati
- → Ne prevelike razlike uz manje zauzeće memorije

Poboljšane slučajne šume imaju potencijala



## Literatura

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Hvala na pažnji!