



Progetto Computational and Statistical Learning

(Dataset CIFAR-10)

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OBIETTIVI

- Descrizione e analisi del dataset **CIFAR-10**
- Descrizione della progettazione (**incrementale**) dei modelli: partire da metodologie semplici come **K-Nearest** per avere una performance di base per poi implementare modelli sempre più complessi come **Random Forest** e **CNN**
- Studiare l'effetto della **PCA** sui modelli
- **Confronto delle performance e dei tempi**

1. Descrizione dataset CIFAR-10

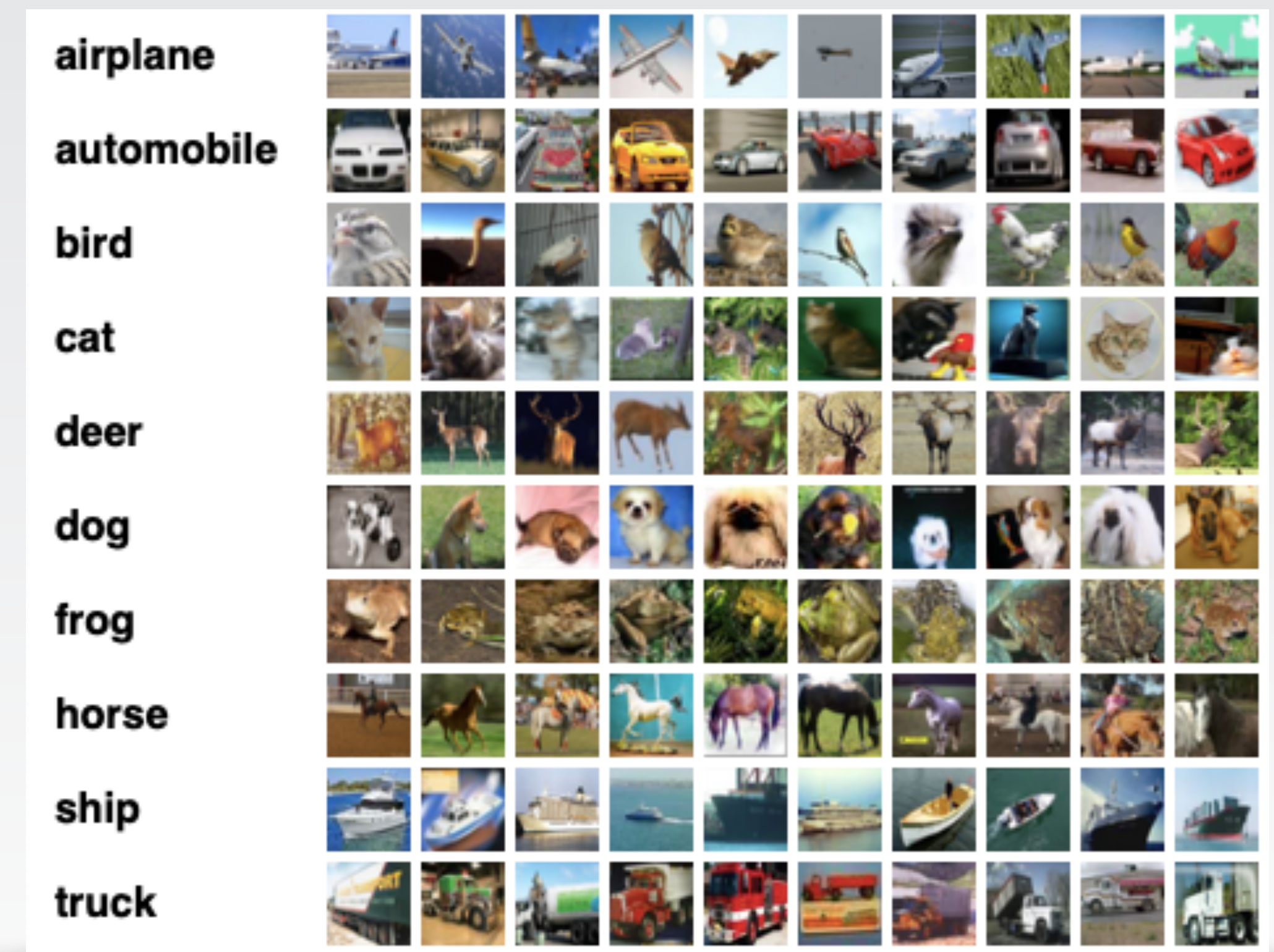
2. Progettazione modelli previsionali

3. Valutazione e confronto modelli

DATASET CIFAR-10

- **60000 immagini** 32x32 pixel a 3 canali (RGB da 0 a 255)
- **10 classi** (mutuamente esclusive) con 6000 immagini per classe
- 50000 immagini di train (5000 immagini per classe) e 10000 immagini di test (1000 immagini per classi)
- **Train e test bilanciati**
- Immagini salvate **row-major** (60000x3072)
- Label codificate con interi da 0 a 9
- Dimensione: **163MB**

OBIETTIVO: creazione e confronto di modelli previsionali in grado di effettuare **classificazione multipla di immagini**.



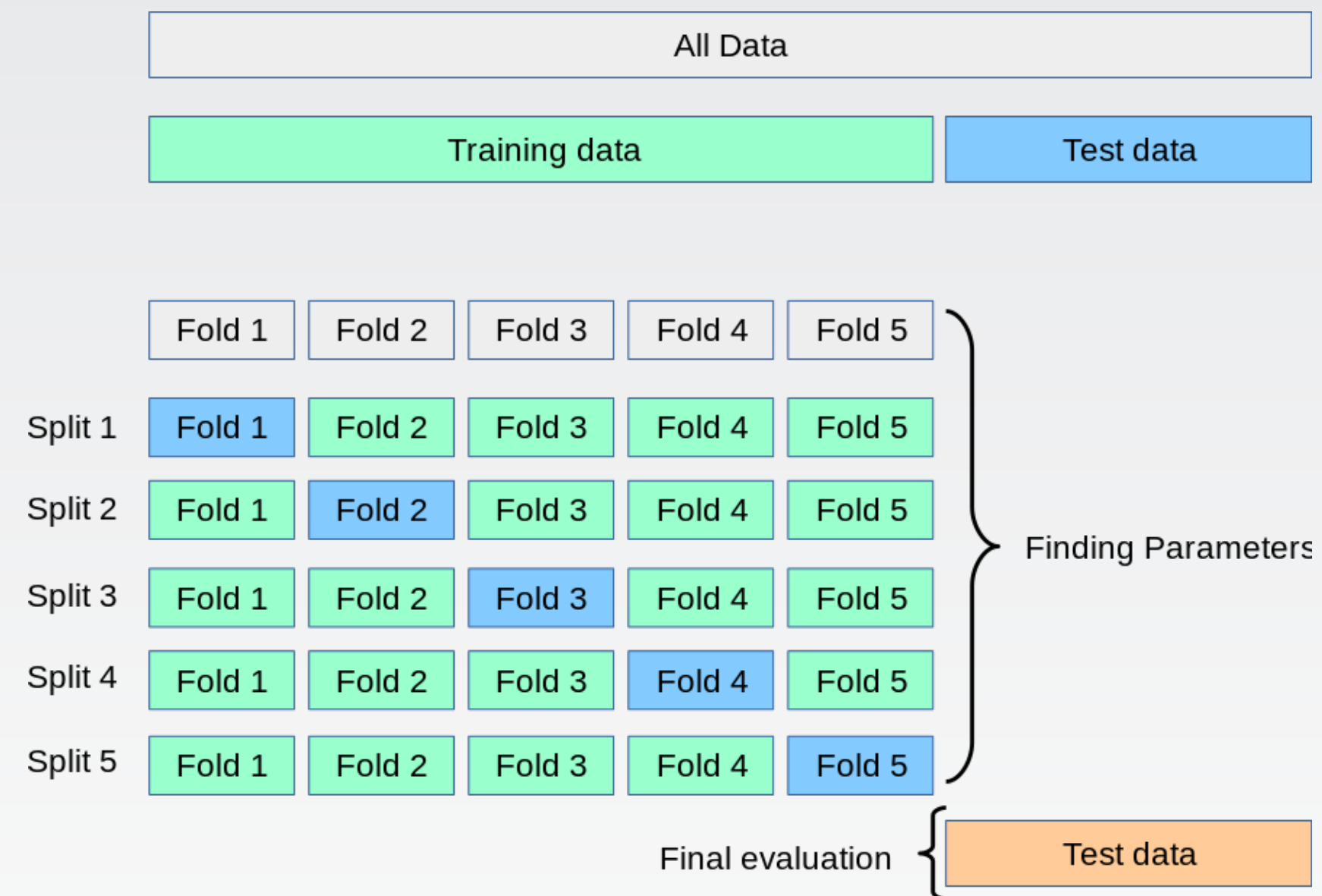
1. Descrizione dataset CIFAR-10

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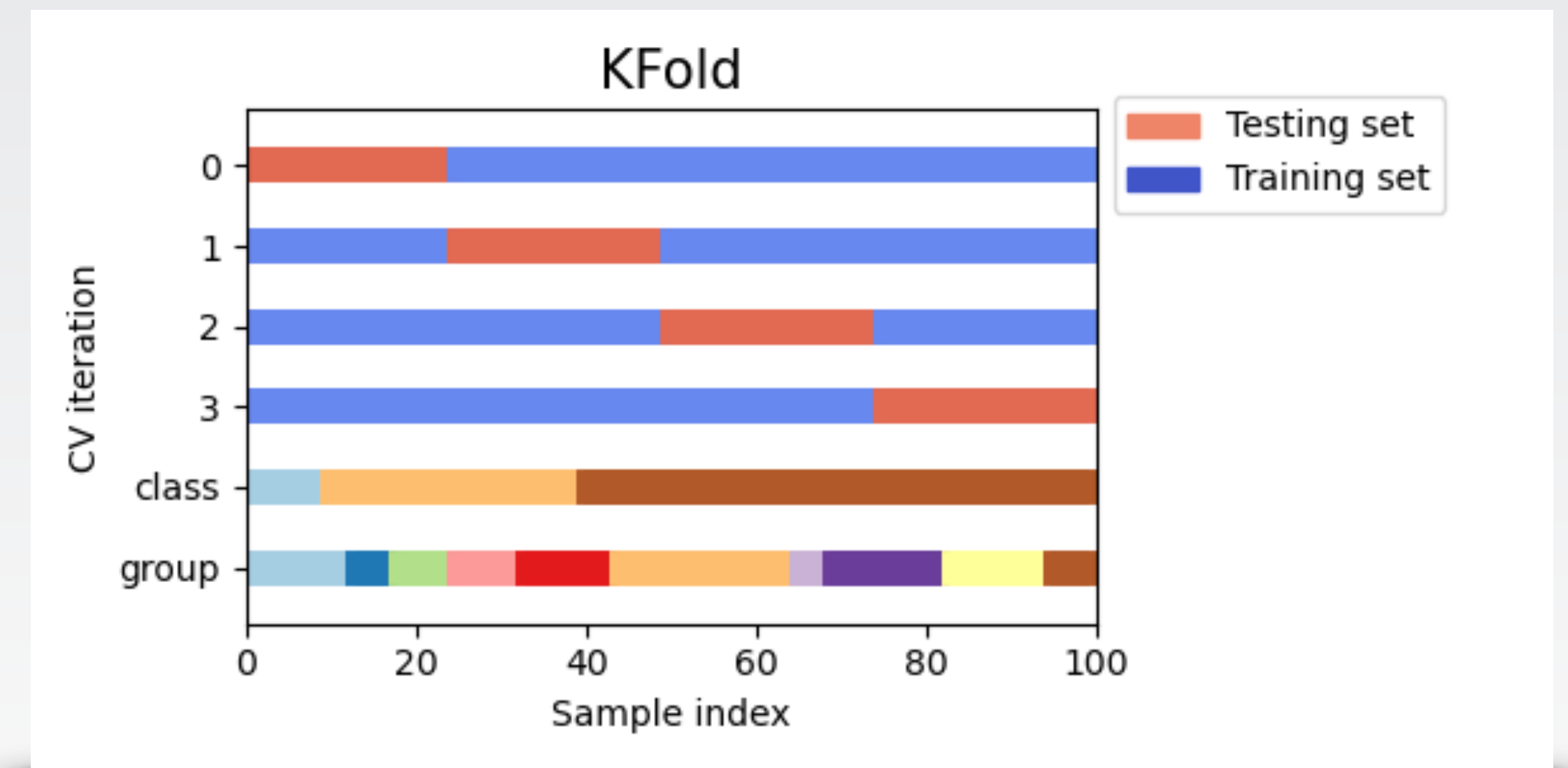
GRID-SEARCH e CROSS-VALIDATION

Grid-Search con **CV** come metodo di ricerca dei parametri migliori:



Metrica = **accuratezza**

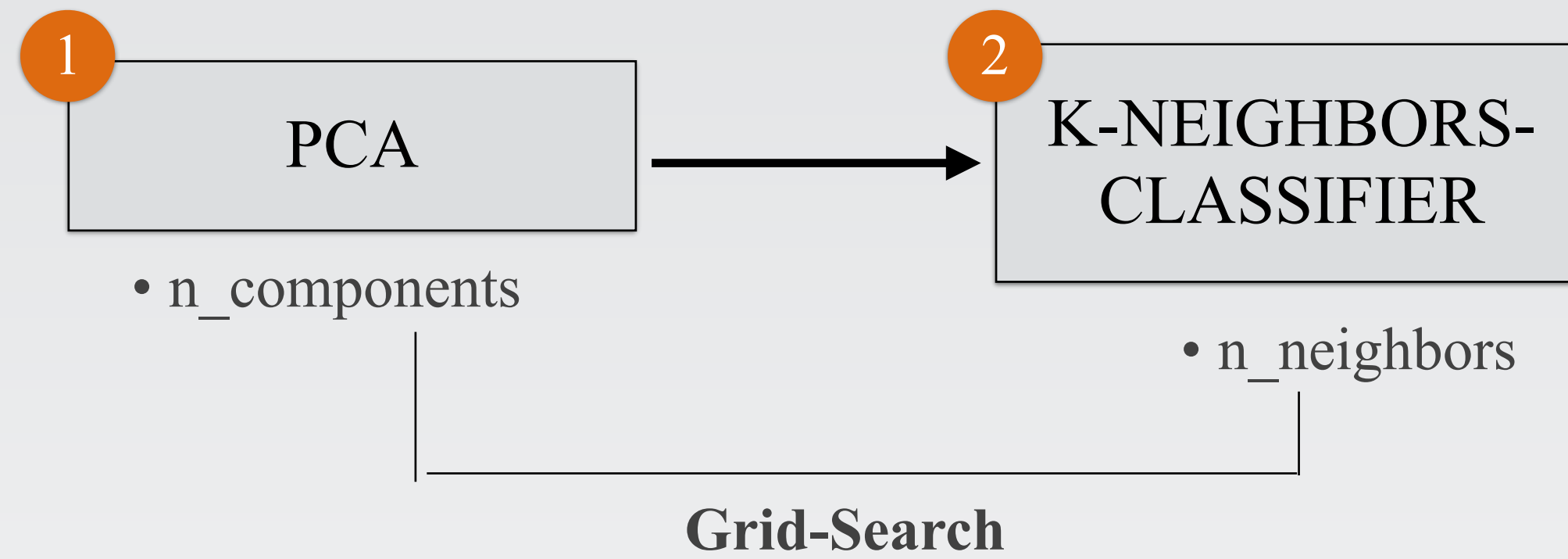
Strategia di **splitting CV** (utilizzo KFold con **K=5**):



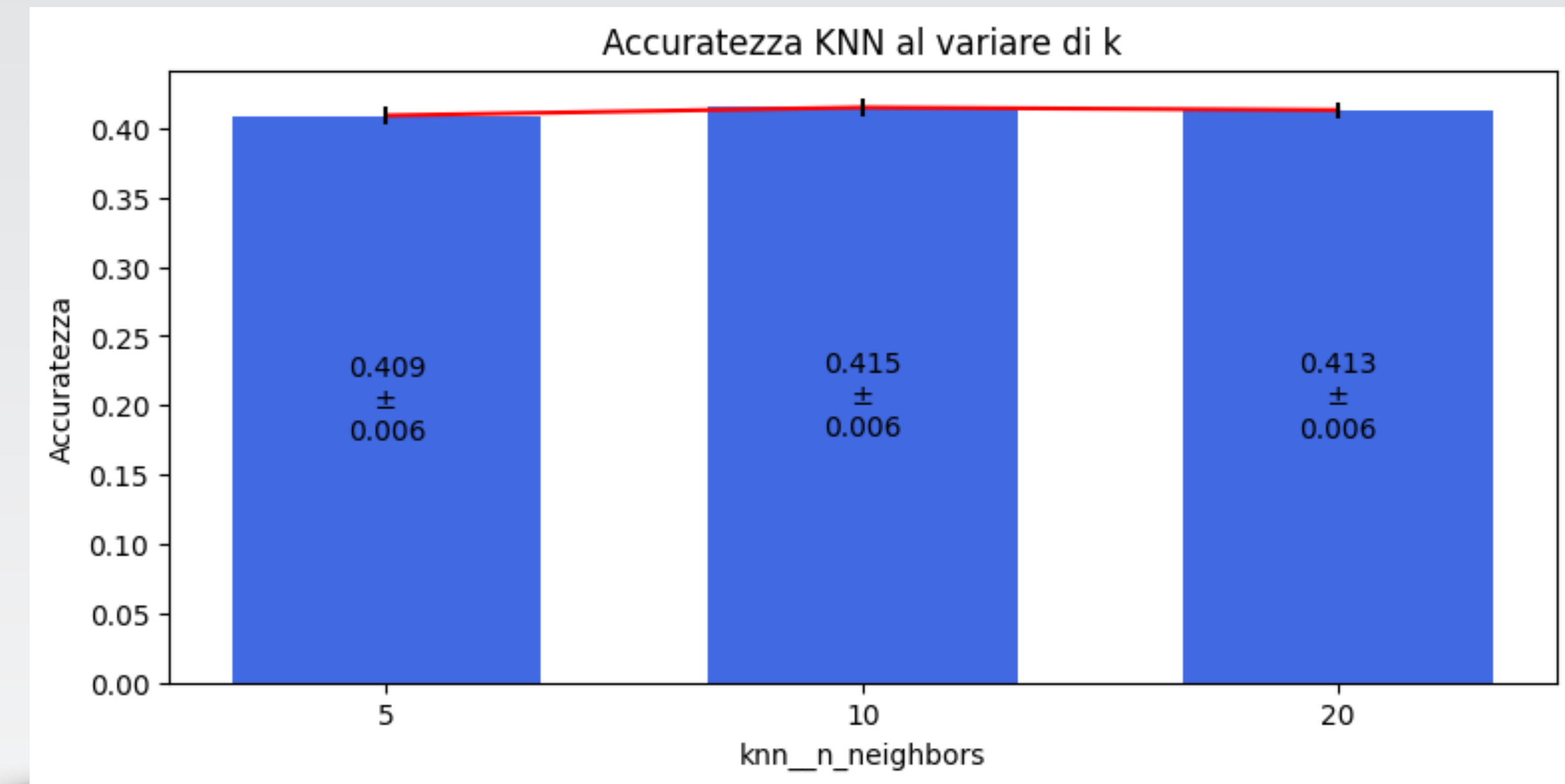
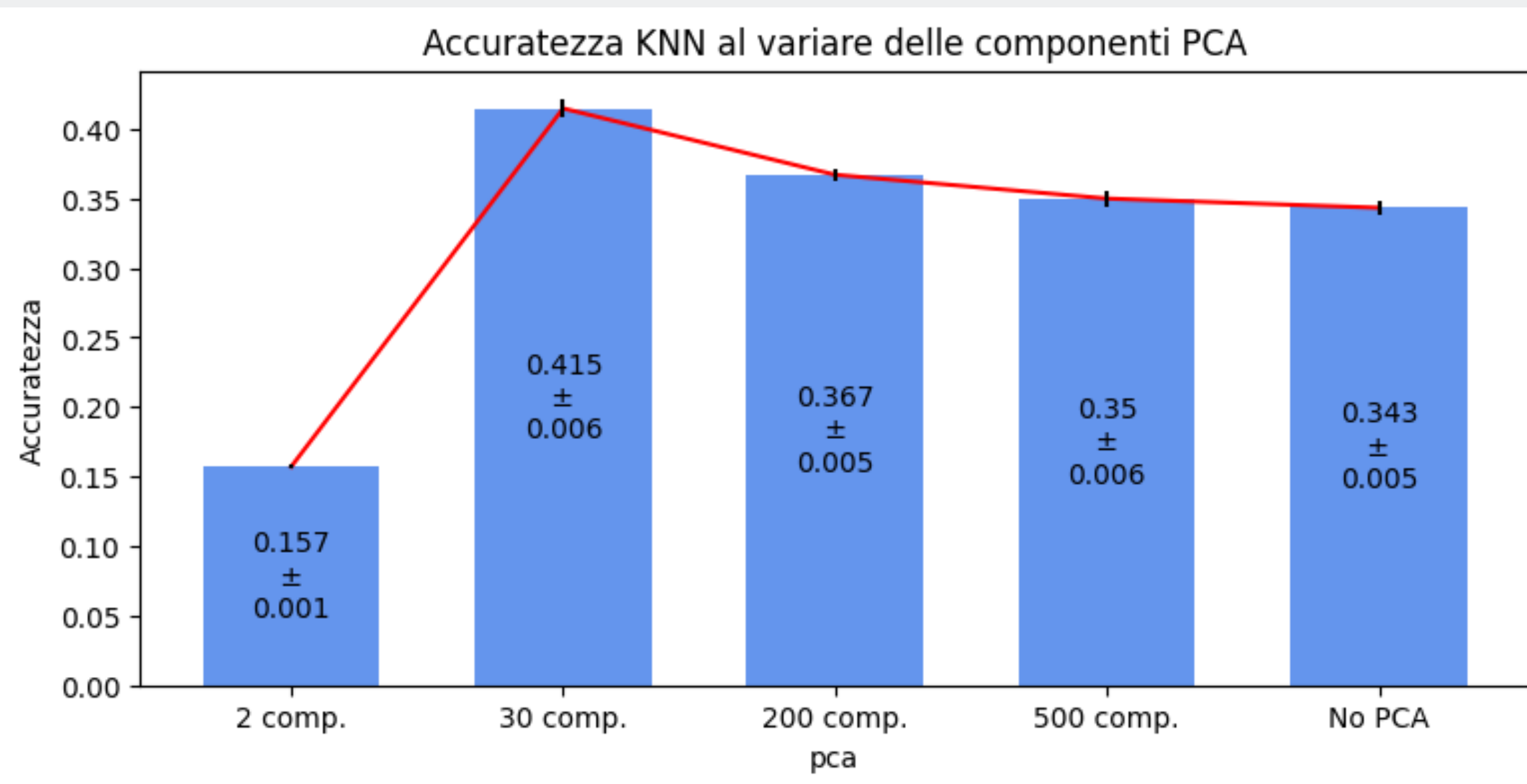
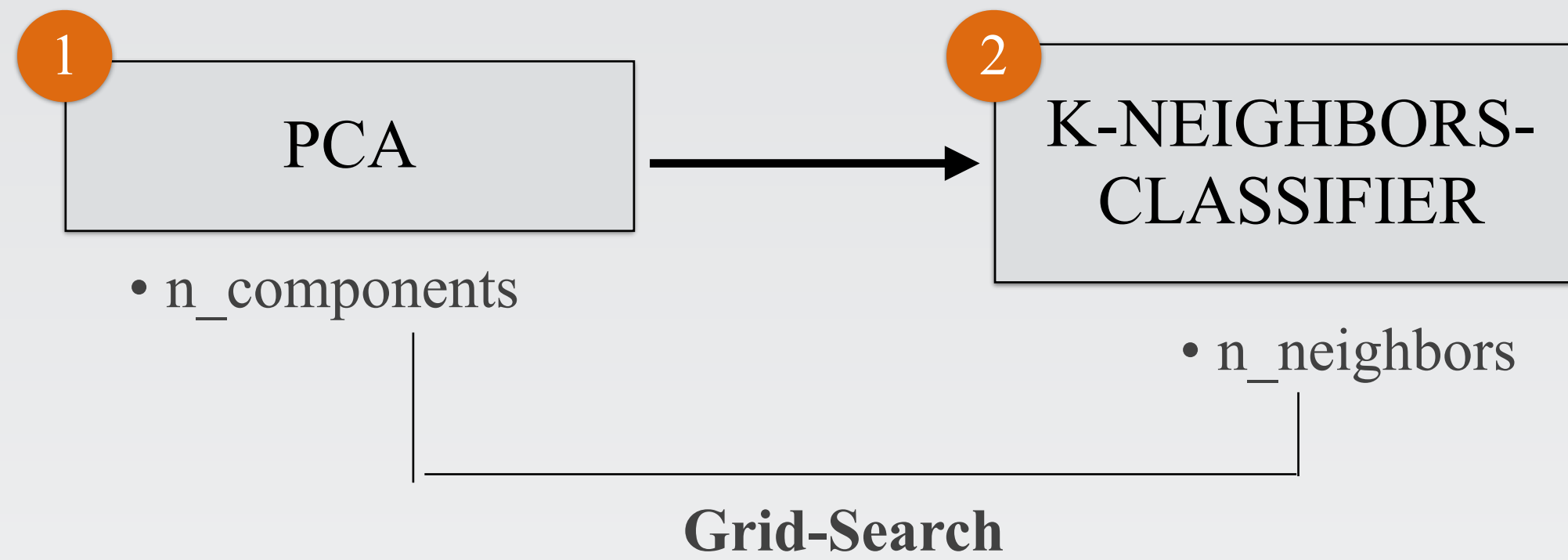
Ad ogni iterazione CV:

- Train size = 40000 immagini
- Test size = 10000 immagini

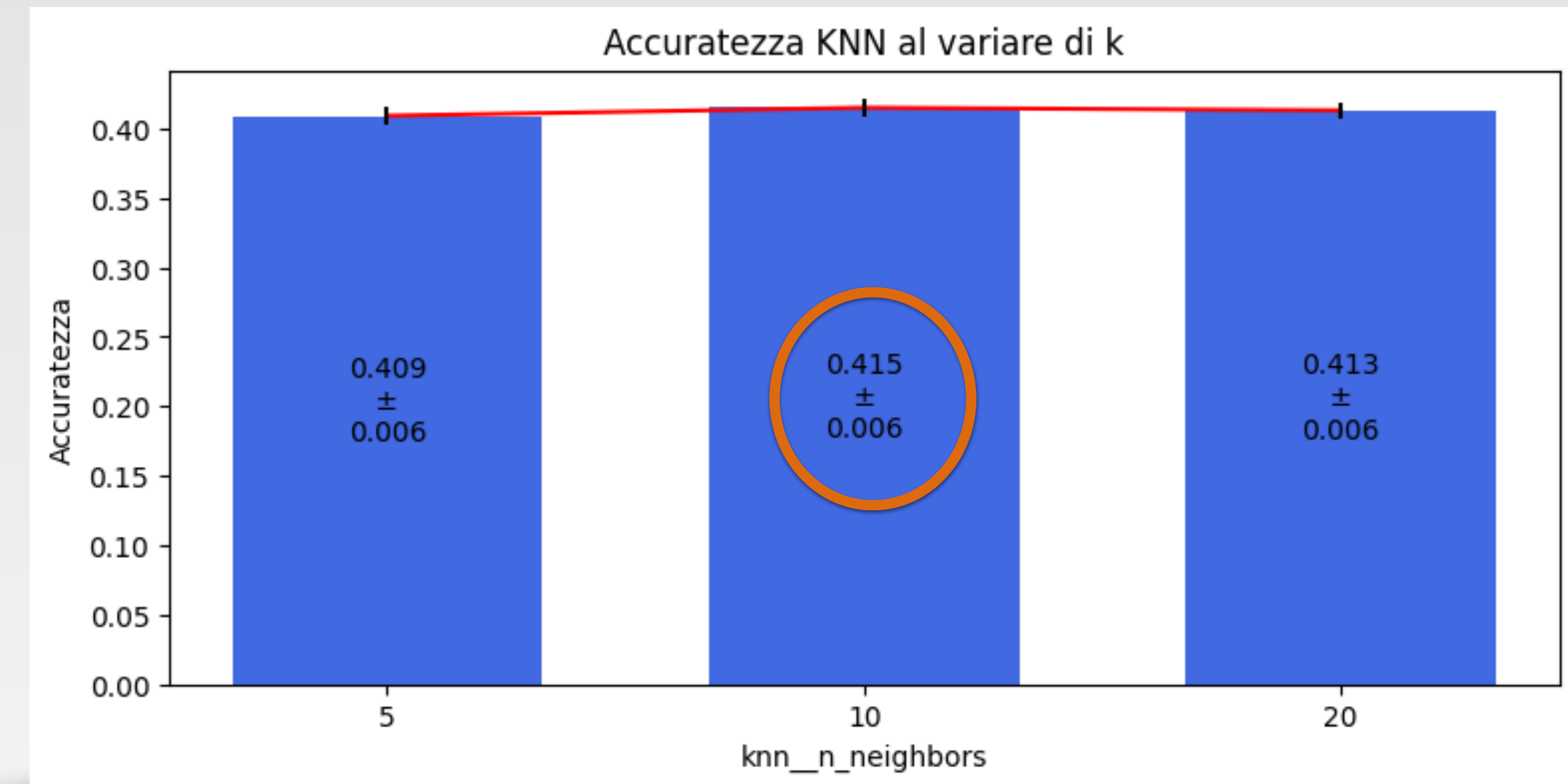
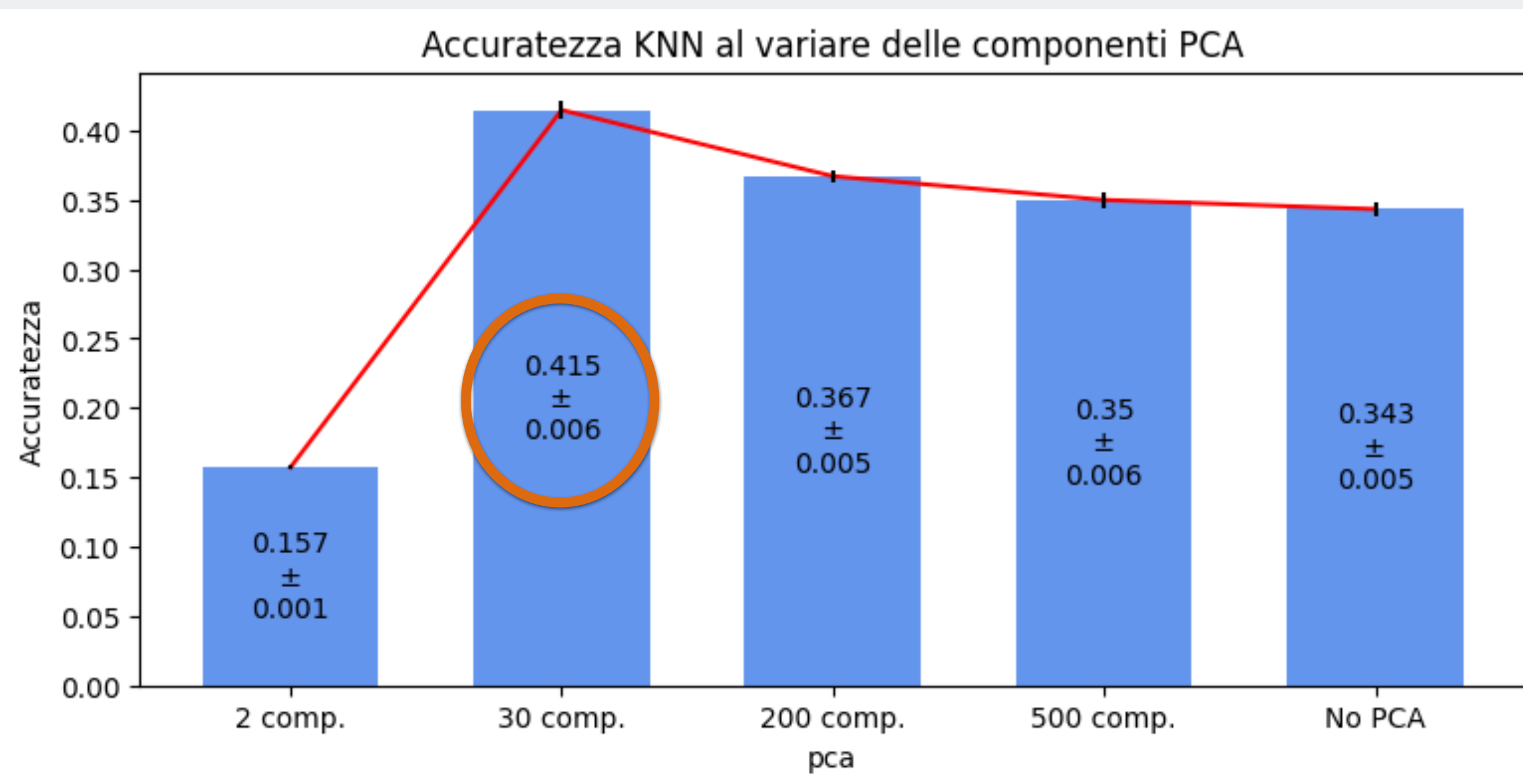
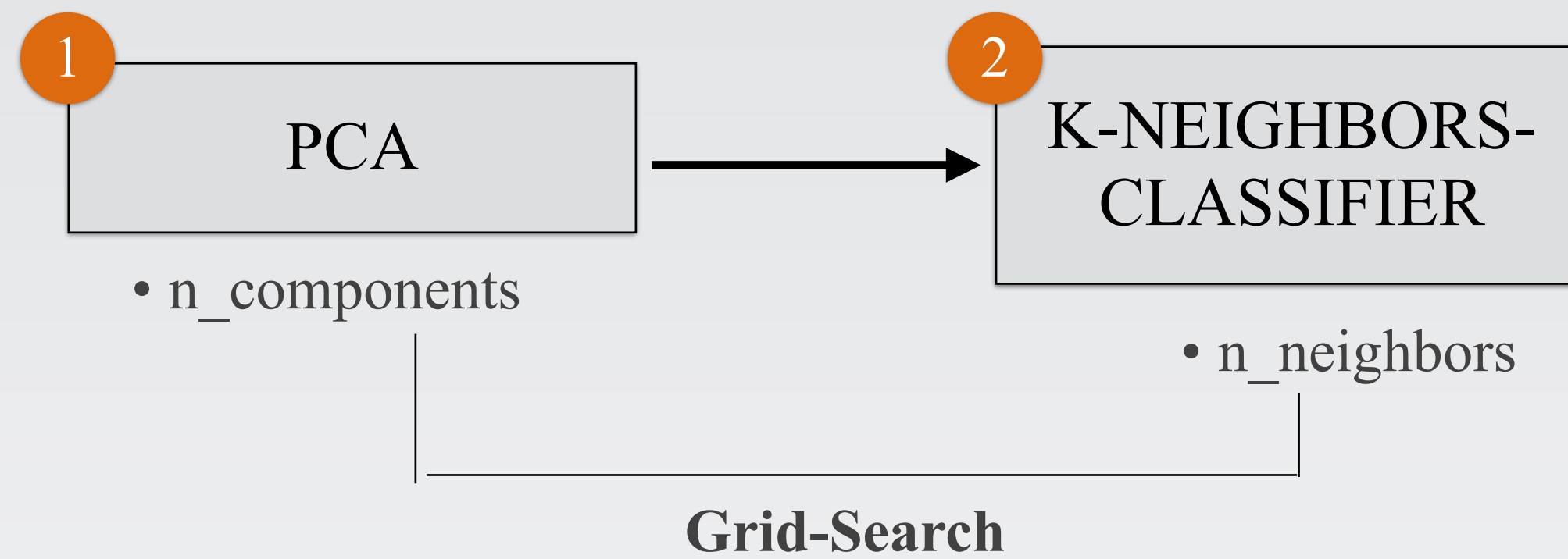
K-NEIGHBORS-CLASSIFIER



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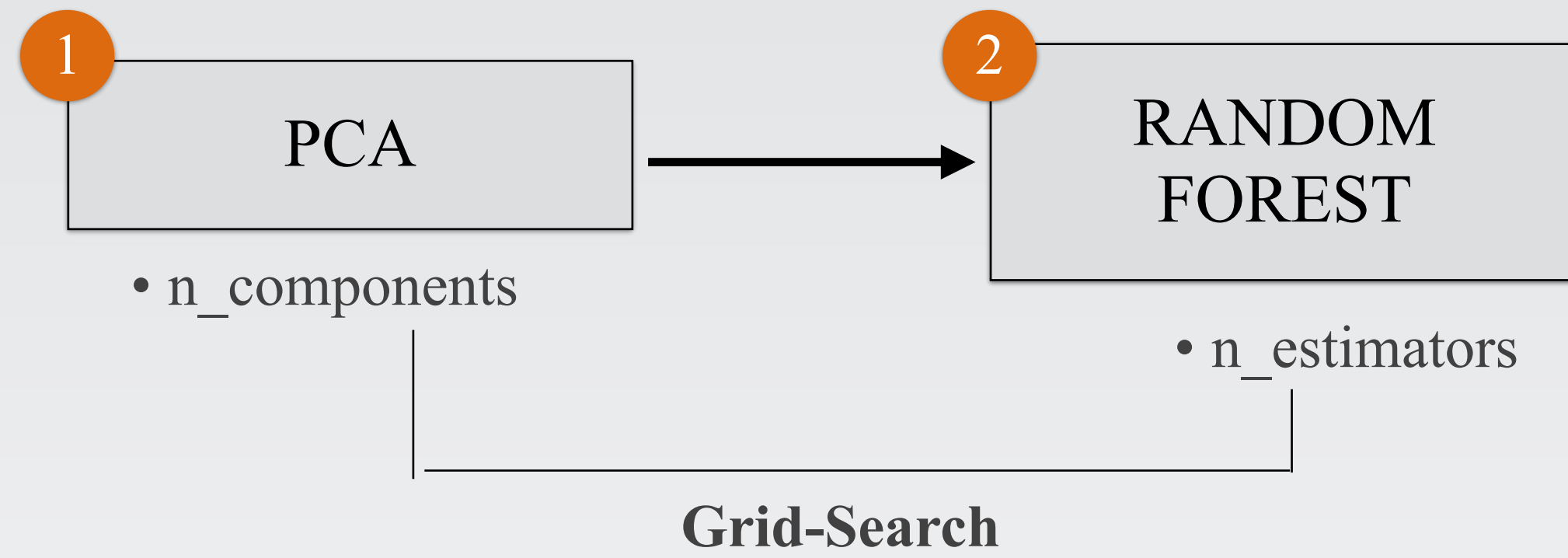


Parametri migliori

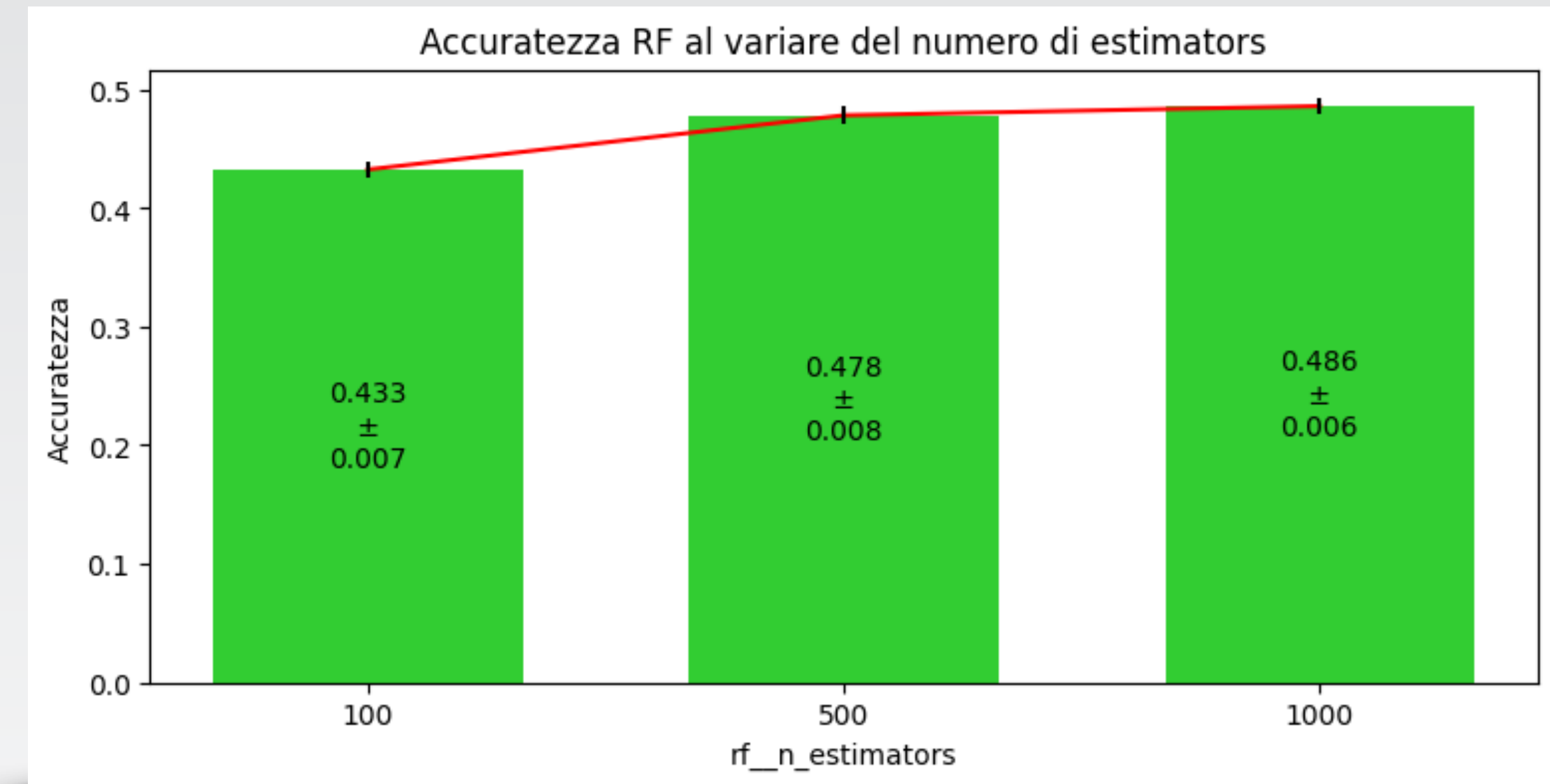
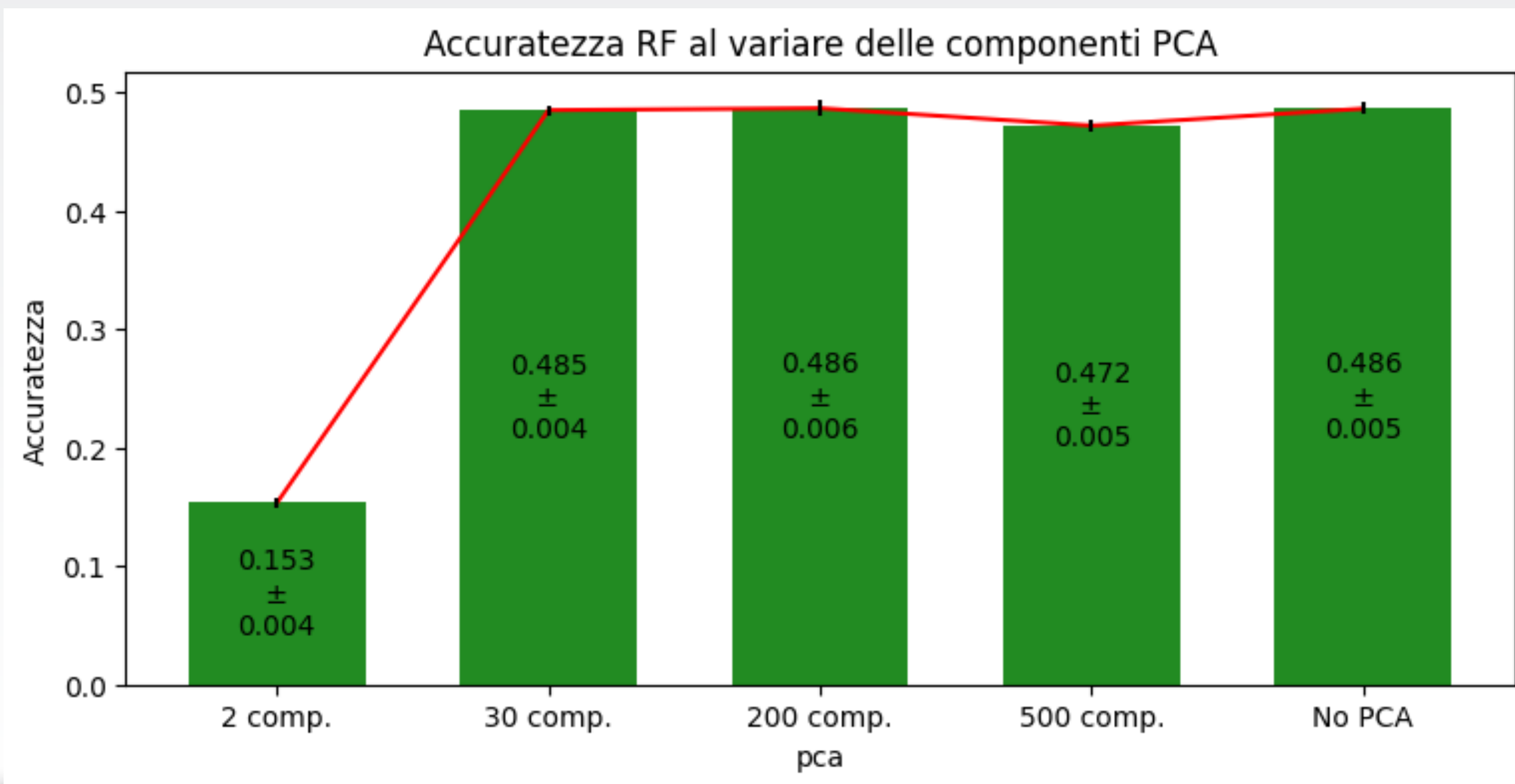
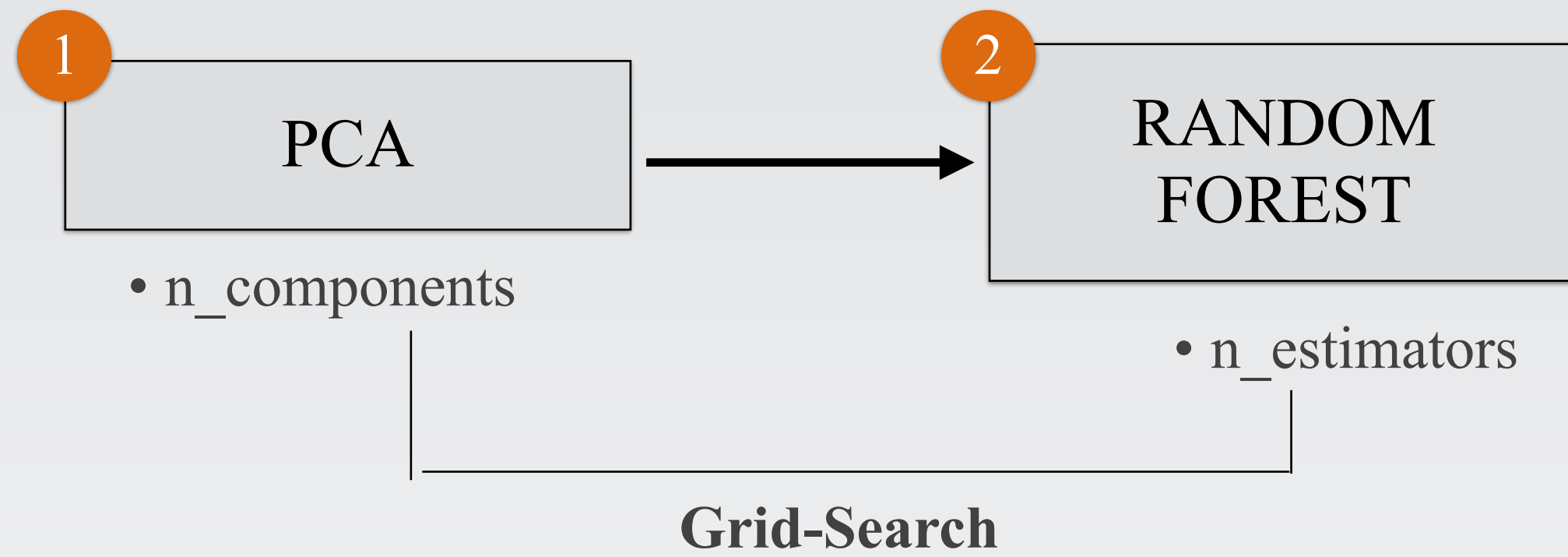
PCA
n_components = 30

KNN
n_neighbors = 10

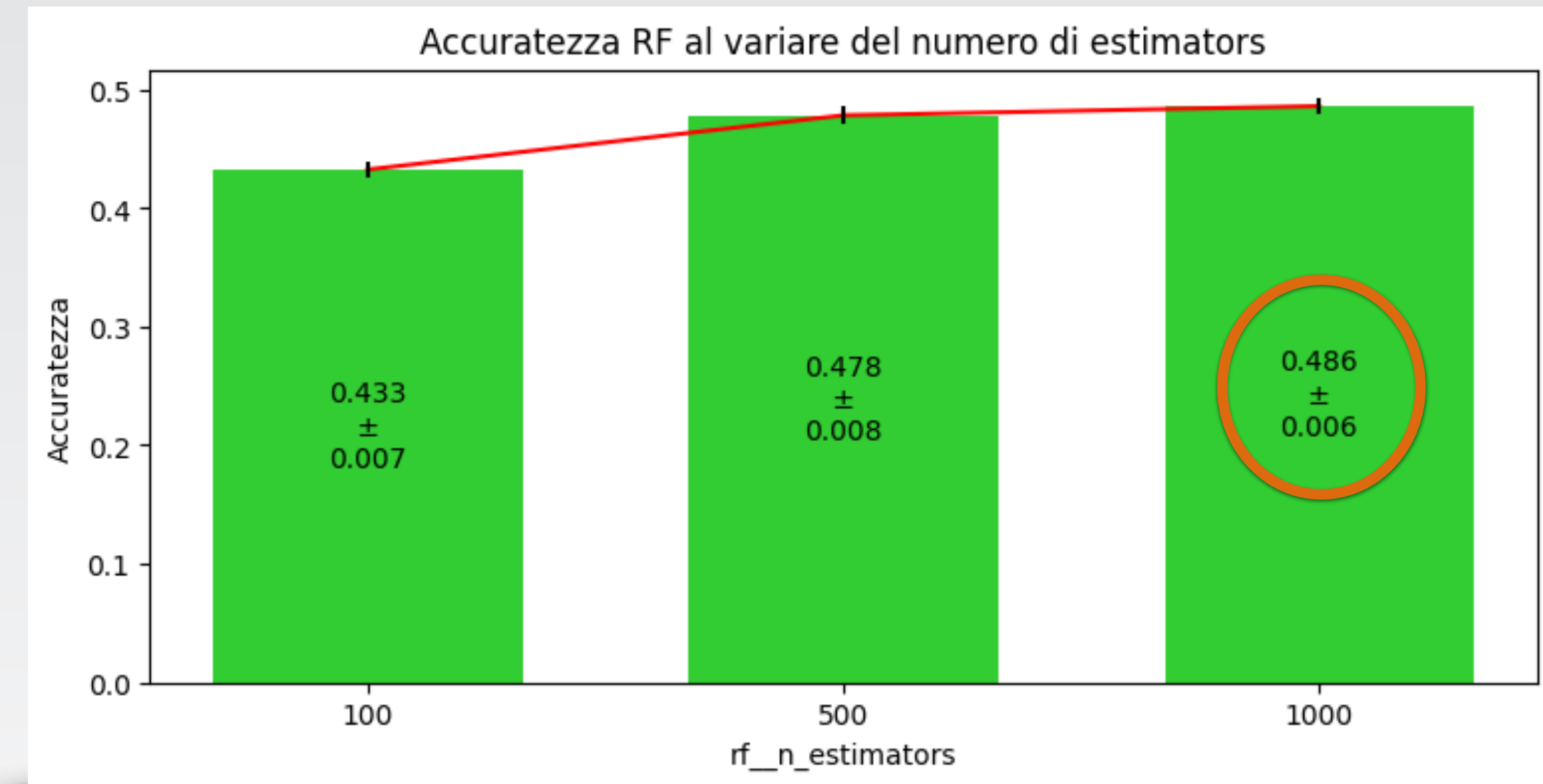
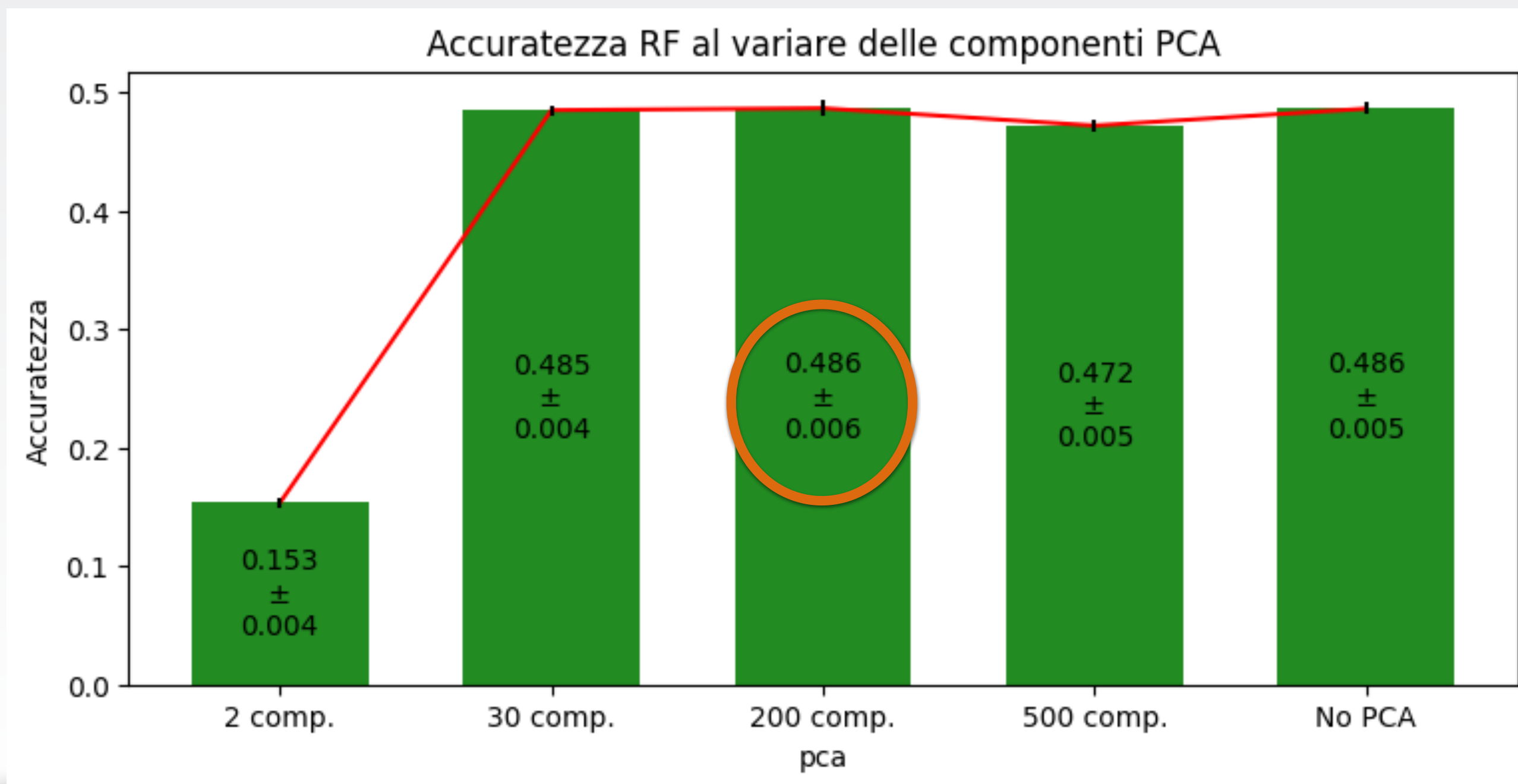
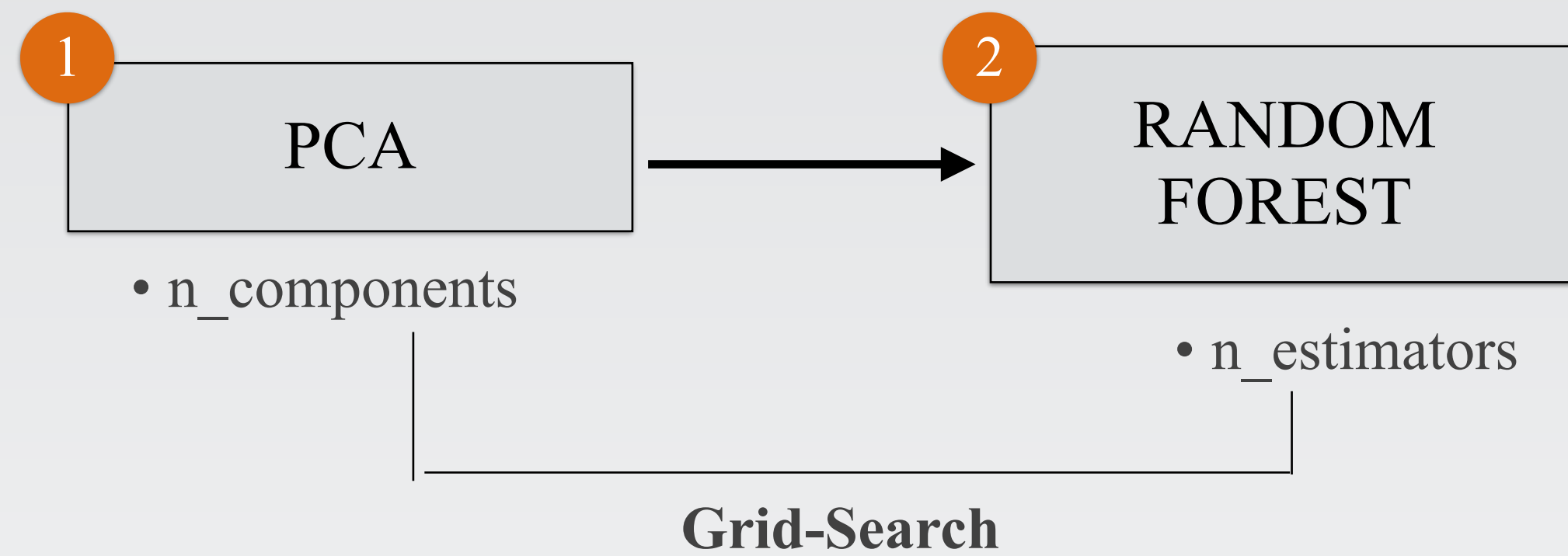
RANDOM FOREST



RANDOM FOREST



RANDOM FOREST



Parametri migliori

PCA	RANDOM FOREST
n_components = 200	n_estimators = 1000

Tempo train: 1029s (200 PCA)
2382s (NO PCA)

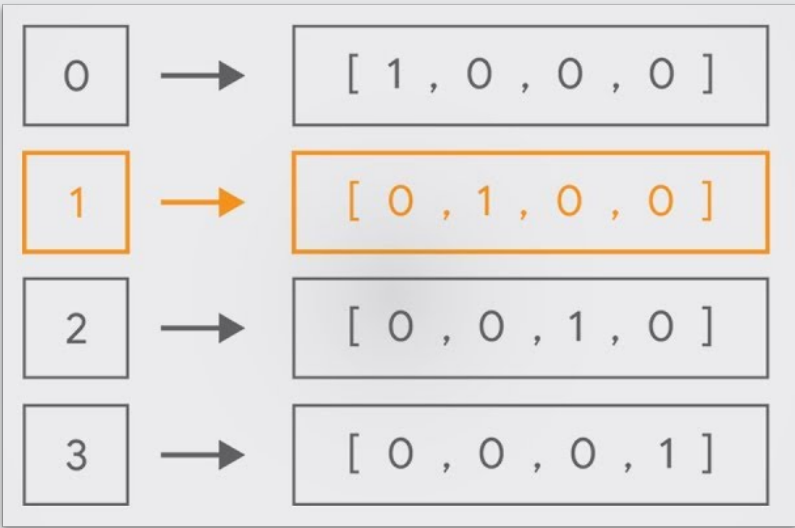
CONVOLUTIONAL NEURAL NETWORK (CNN)

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 32, 32, 32)	896
conv2d_1 (Conv2D)	(None, 32, 32, 32)	9248
max_pooling2d (MaxPooling2D)	(None, 16, 16, 32)	0
dropout (Dropout)	(None, 16, 16, 32)	0
conv2d_2 (Conv2D)	(None, 16, 16, 64)	18496
conv2d_3 (Conv2D)	(None, 16, 16, 64)	36928
max_pooling2d_1 (MaxPooling2D)	(None, 8, 8, 64)	0
dropout_1 (Dropout)	(None, 8, 8, 64)	0
flatten (Flatten)	(None, 4096)	0
dense (Dense)	(None, 128)	524416
dropout_2 (Dropout)	(None, 128)	0
dense_1 (Dense)	(None, 10)	1290

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Total params: 591,274
Trainable params: 591,274
Non-trainable params: 0

One-Hot-Enconding label



Parametri Rete :

4 Layer Convolutivi:

- Kernel 3x3
- Funzione attivazione: ReLu
- Padding “same”
- Num. filtri: 32 e 64

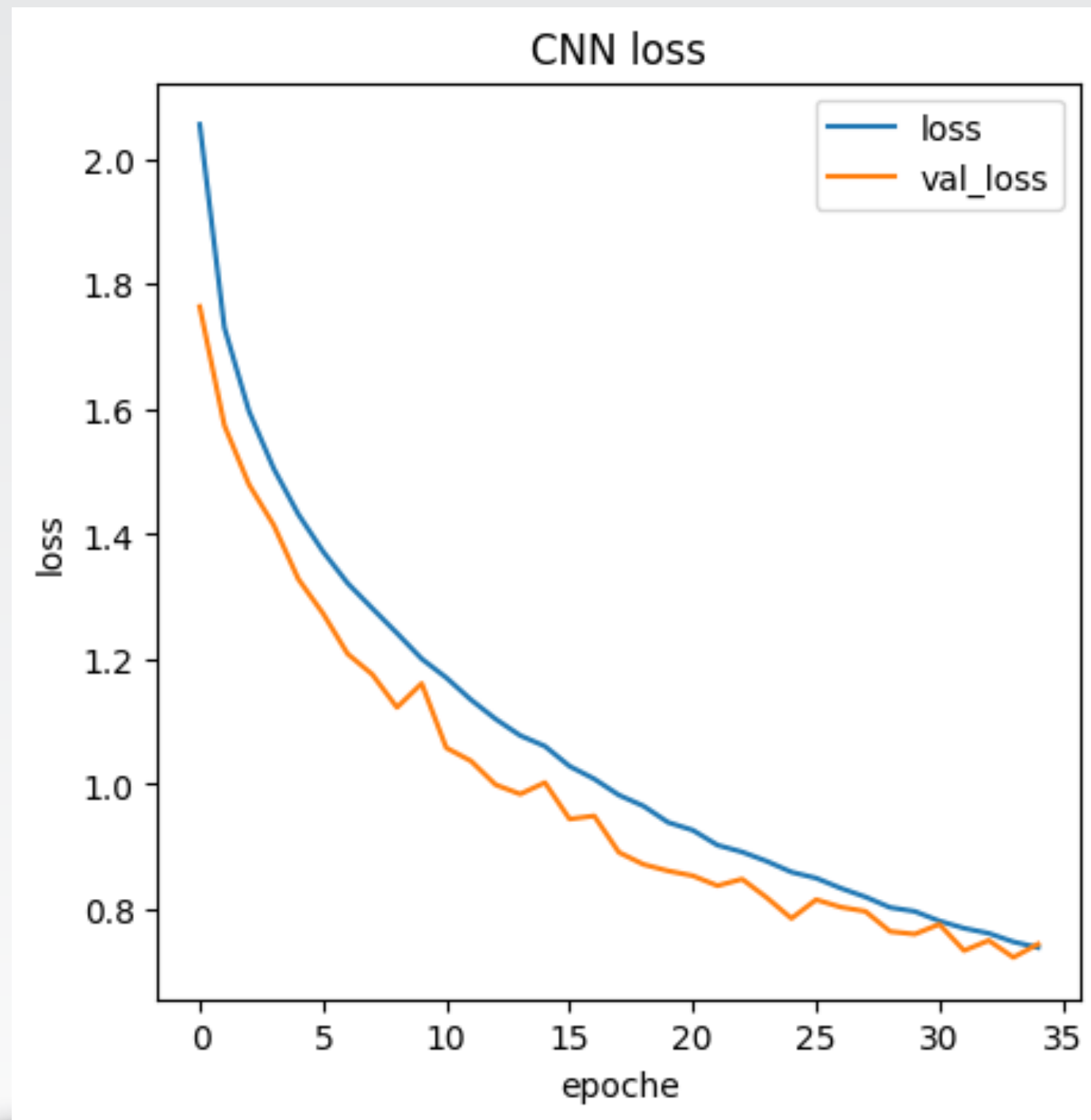
Fully Connected:

- 128 nodi (ReLu)
- 10 nodi (SoftMax)

MaxPooling = 2x2

Dropout = 0.3

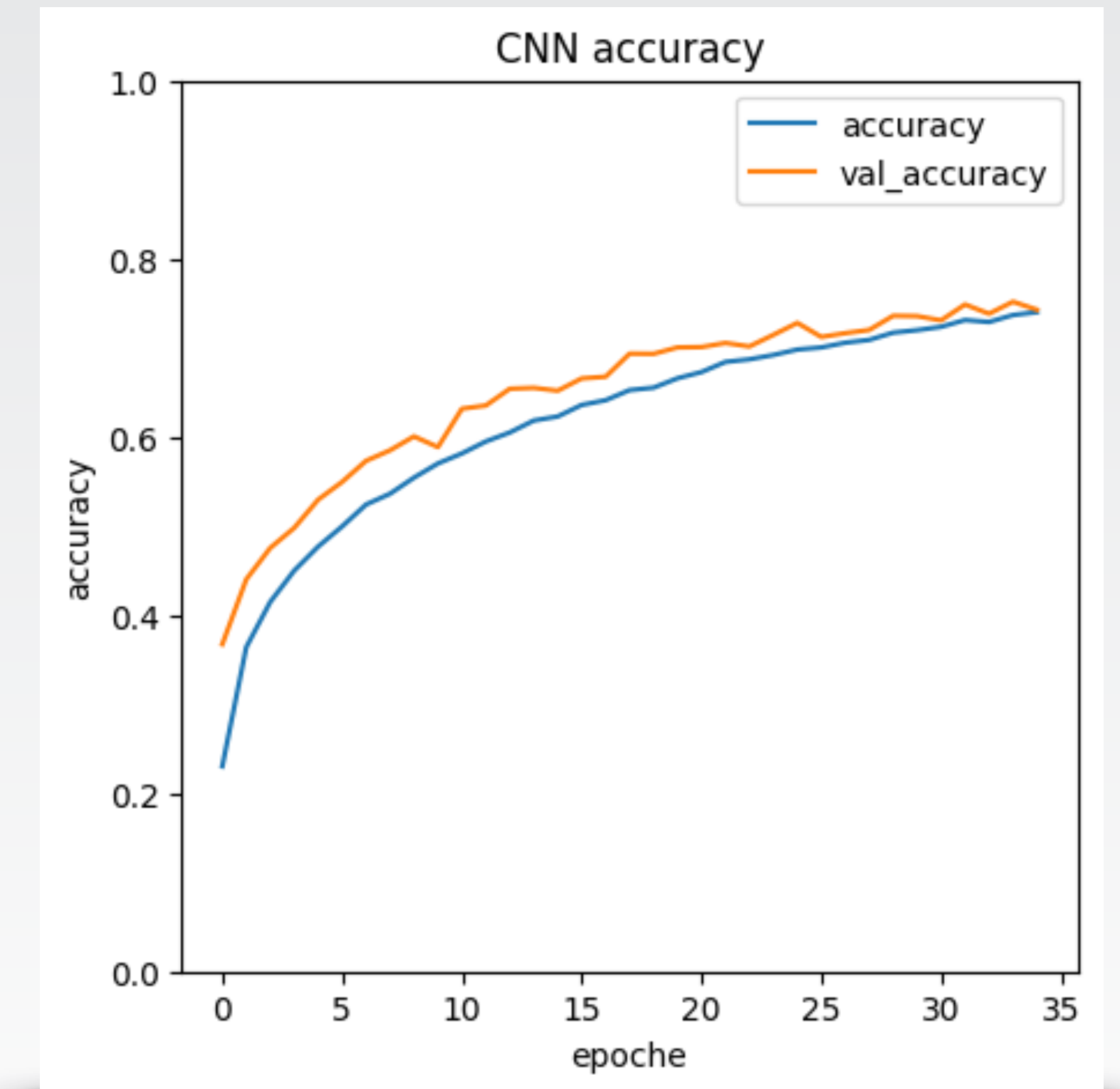
CNN TRAIN



Parametri train:

- Ottimizzatore: **SGD**
- Loss: **categorical_crossentropy**
- Metrica: **accuratezza**
- Epoche: **35**
- Batch size: **64**

Tempo train totale: **5423s**



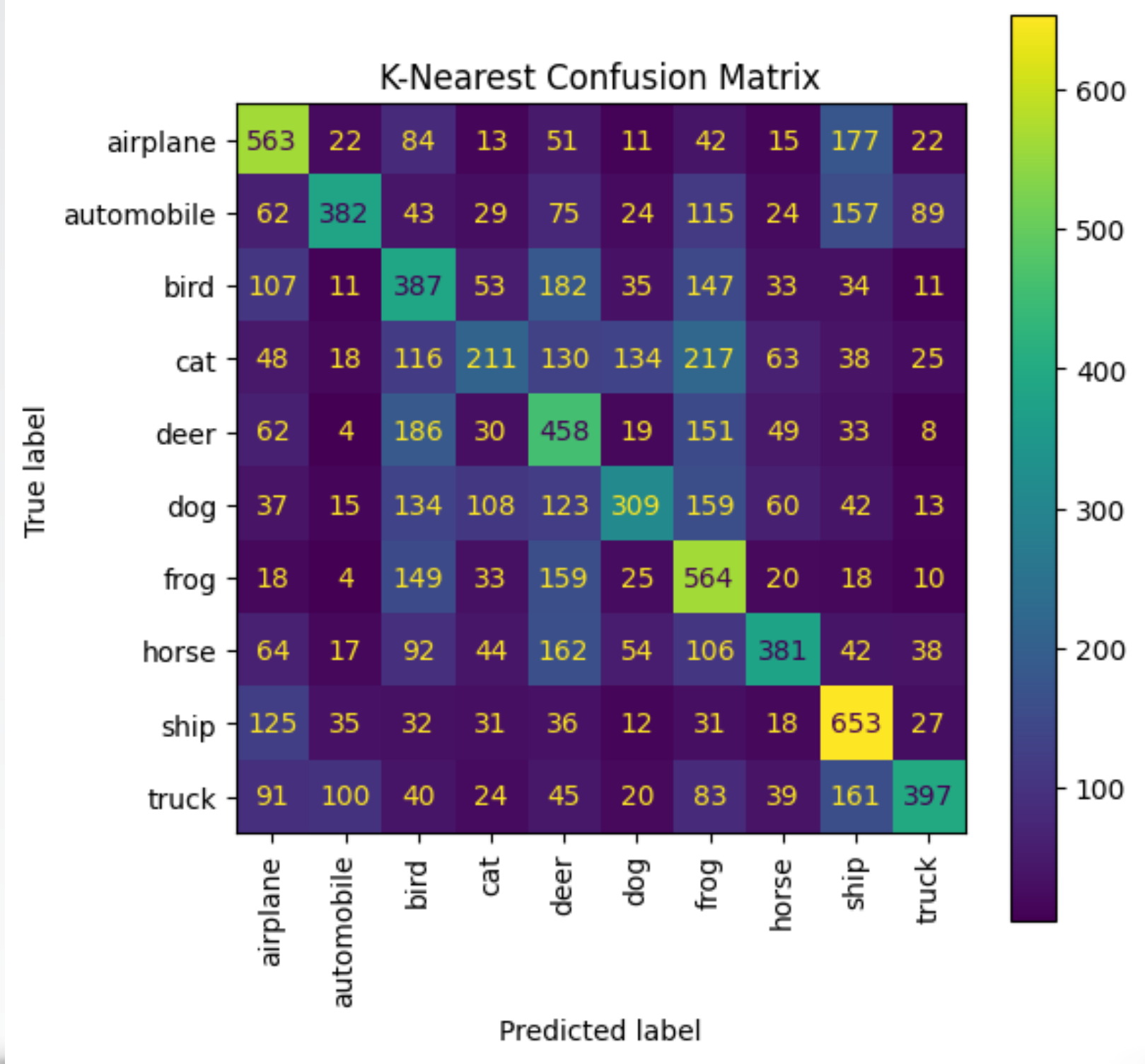
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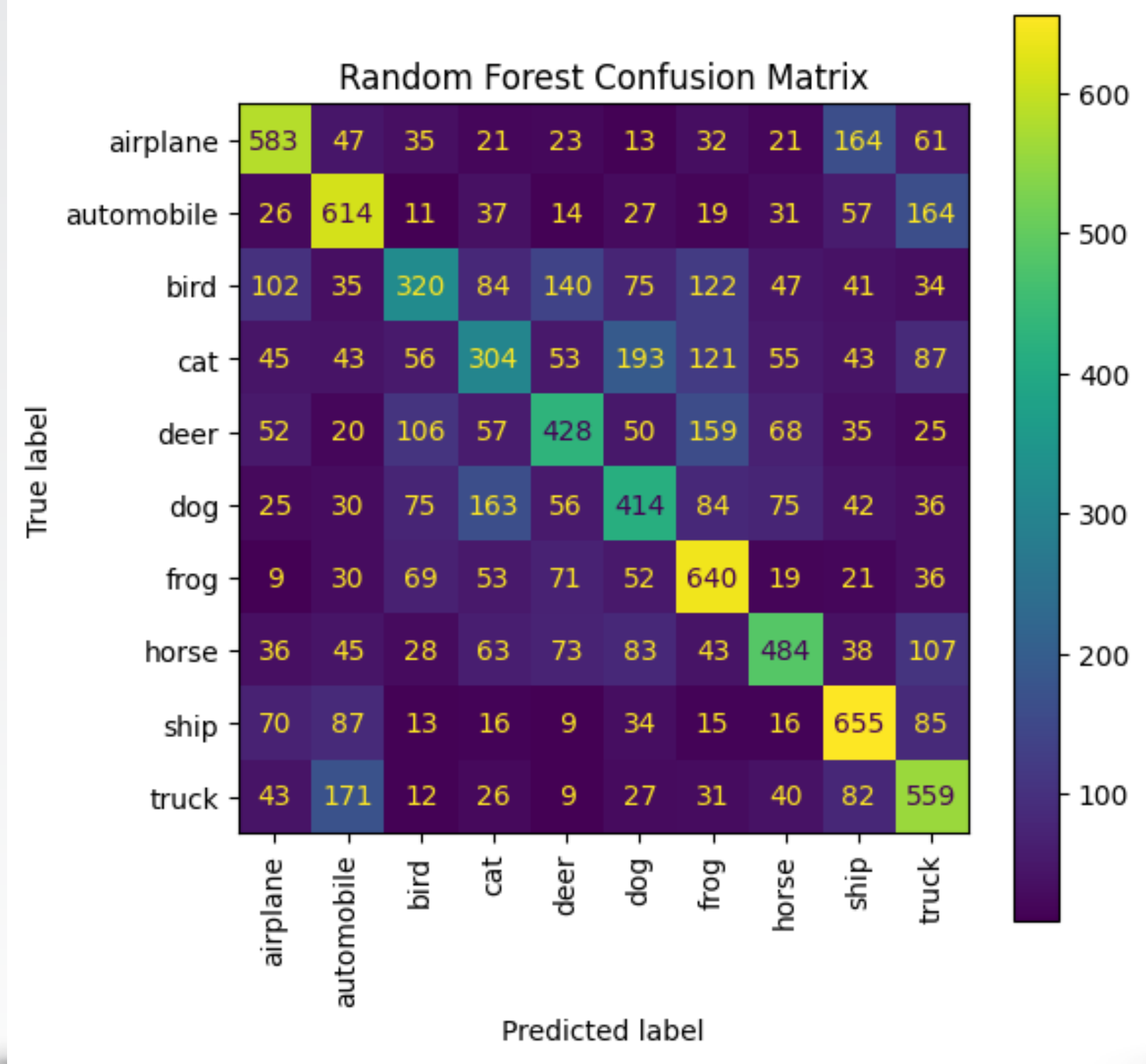
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CONFUSION MATRIX

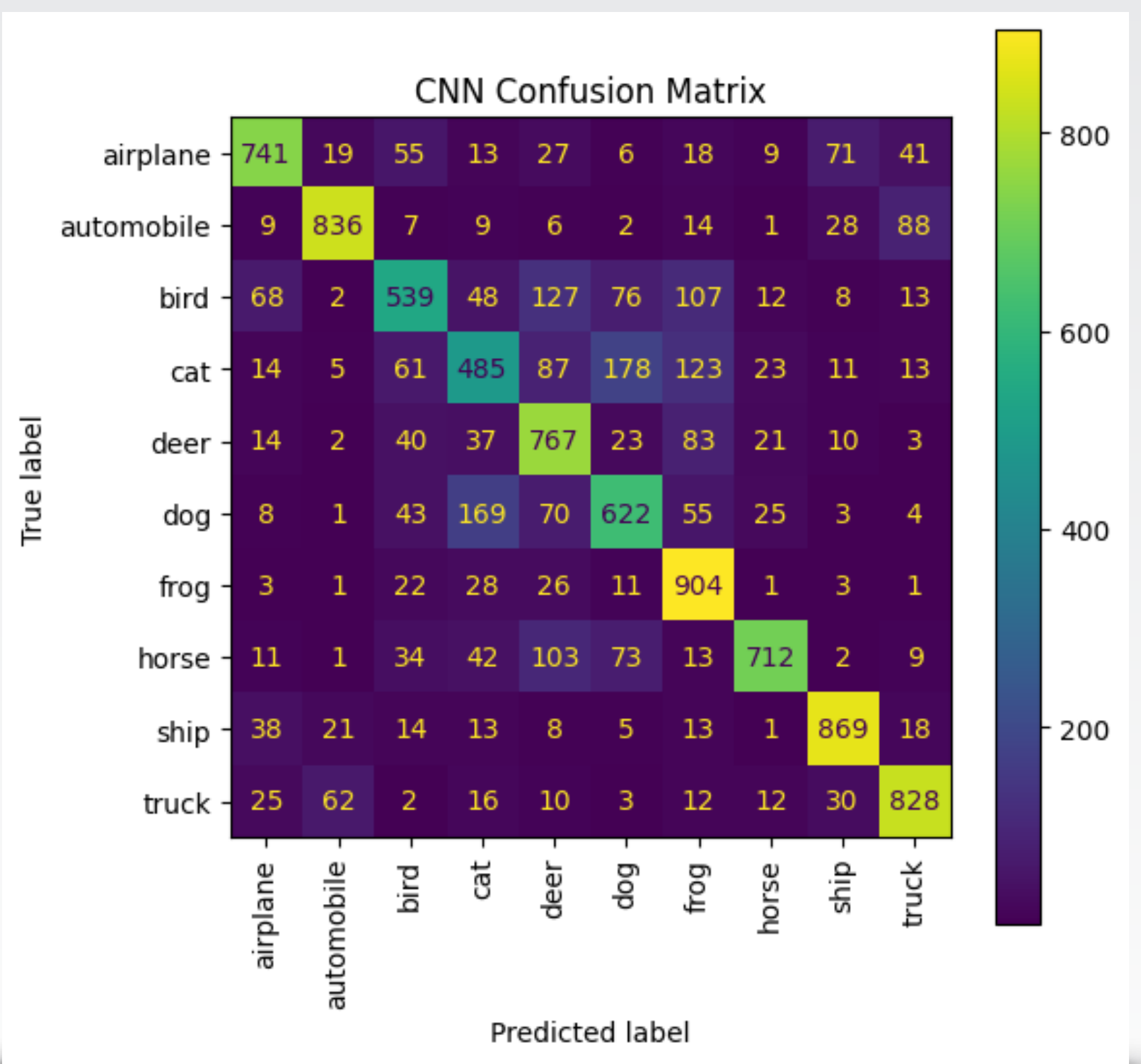
KNN



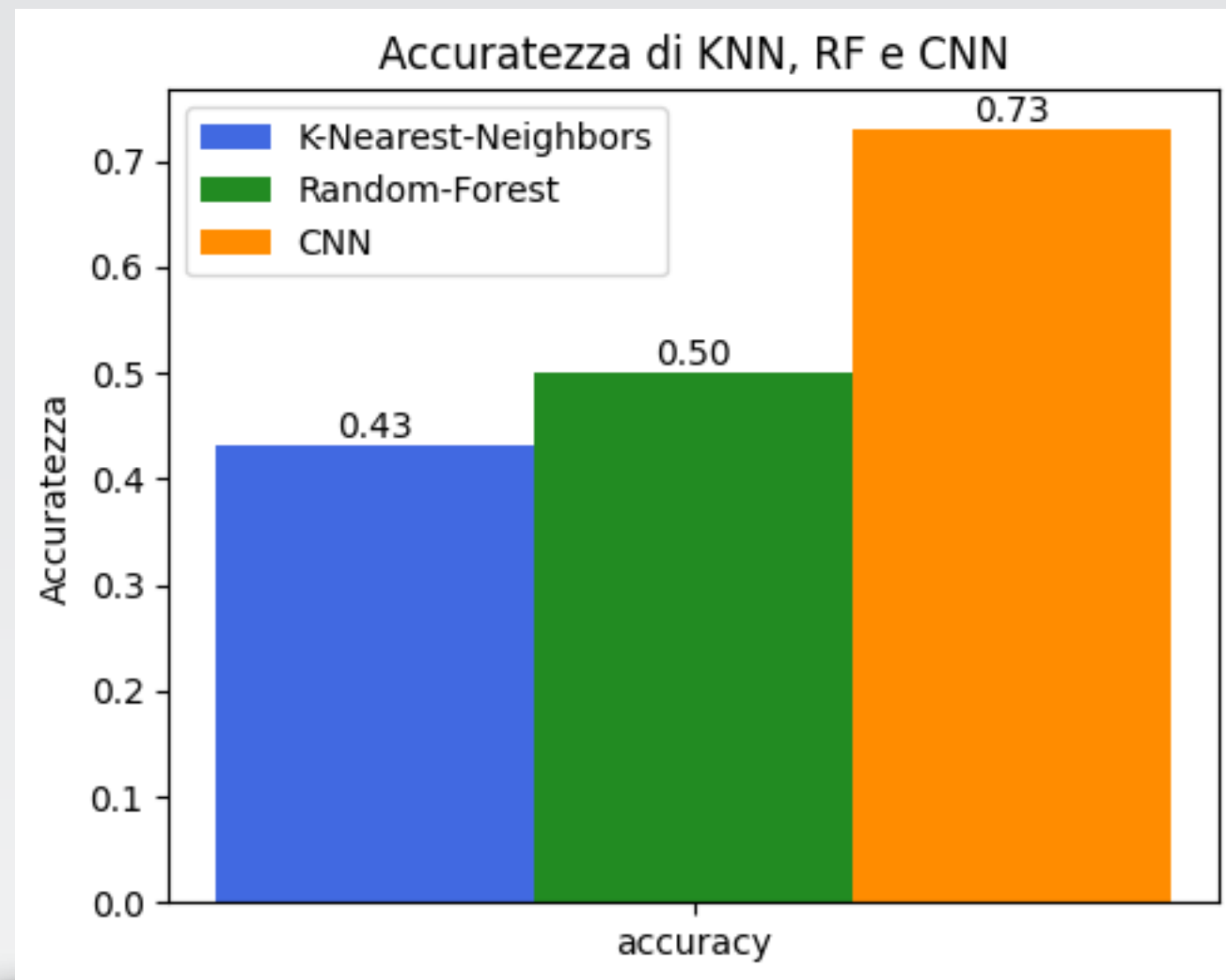
Random Forest



CNN

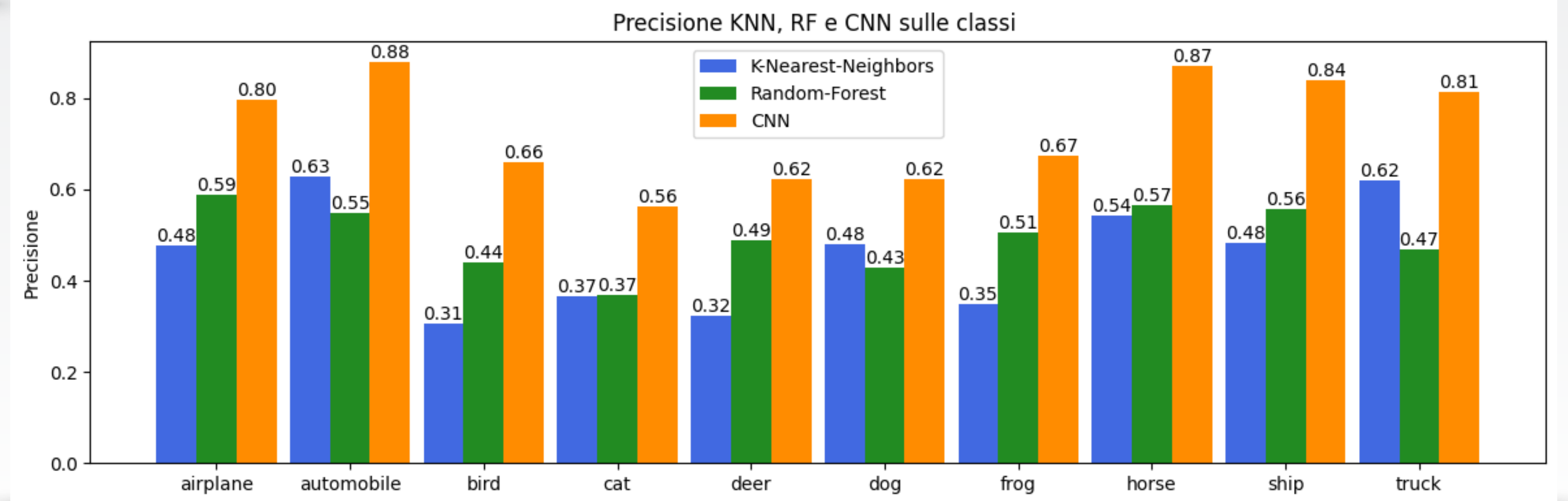


ACCURATEZZA

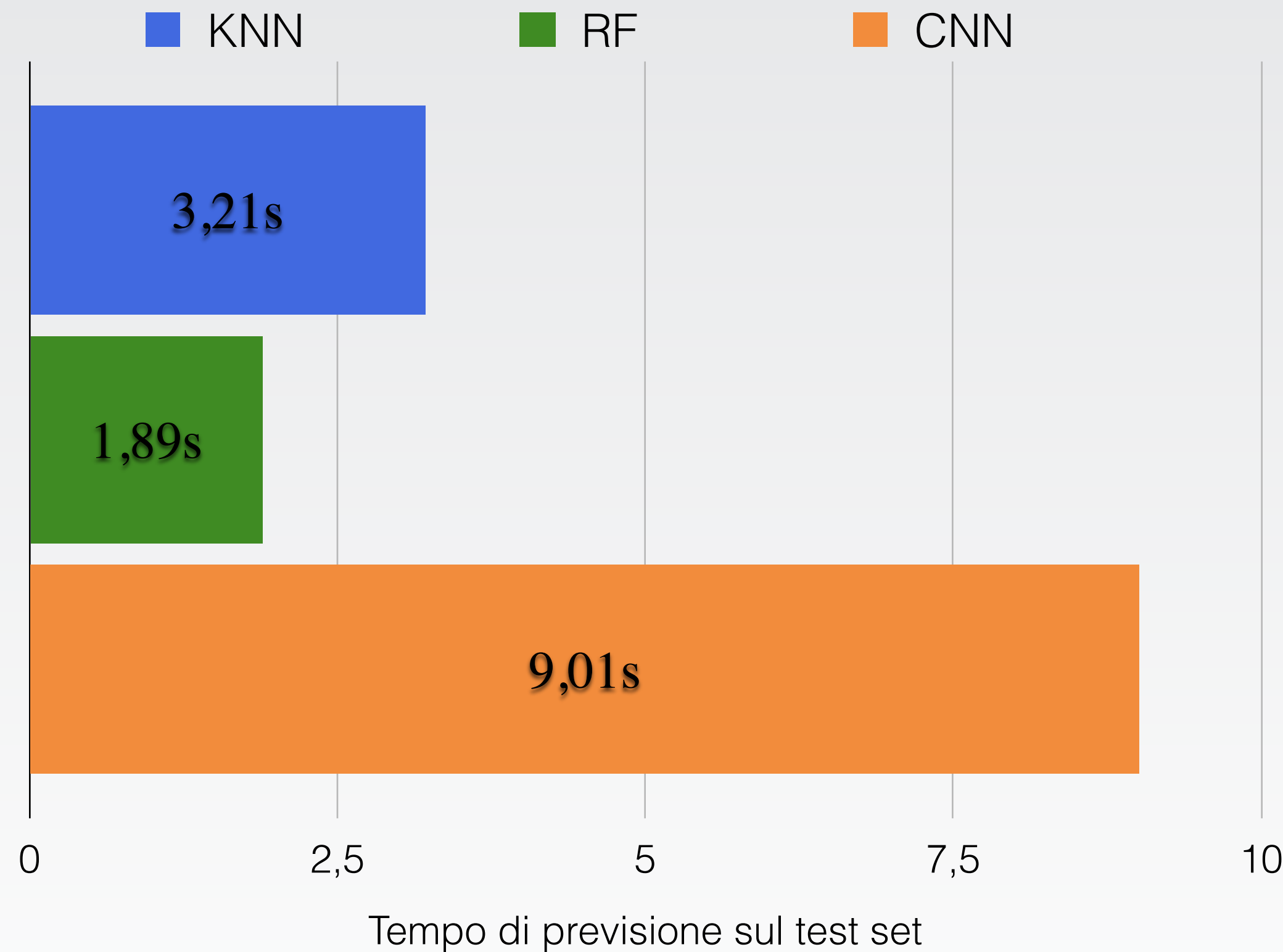


CNN ha un'accuratezza considerevolmente maggiore rispetto a KNN e RF

CNN ha precisione maggiore per tutte le classi mentre RF non sempre è migliore di KNN



TEMPI DI PREVISIONE



MacBook Air

CPU : Intel Core 1,8 GHz i5 dual-core

RAM : 8 GB

- CNN ha accuratezza migliore ma impiega più tempo per la previsione
- RF è il modello con tempi di previsione migliori

Grazie per l'attenzione
