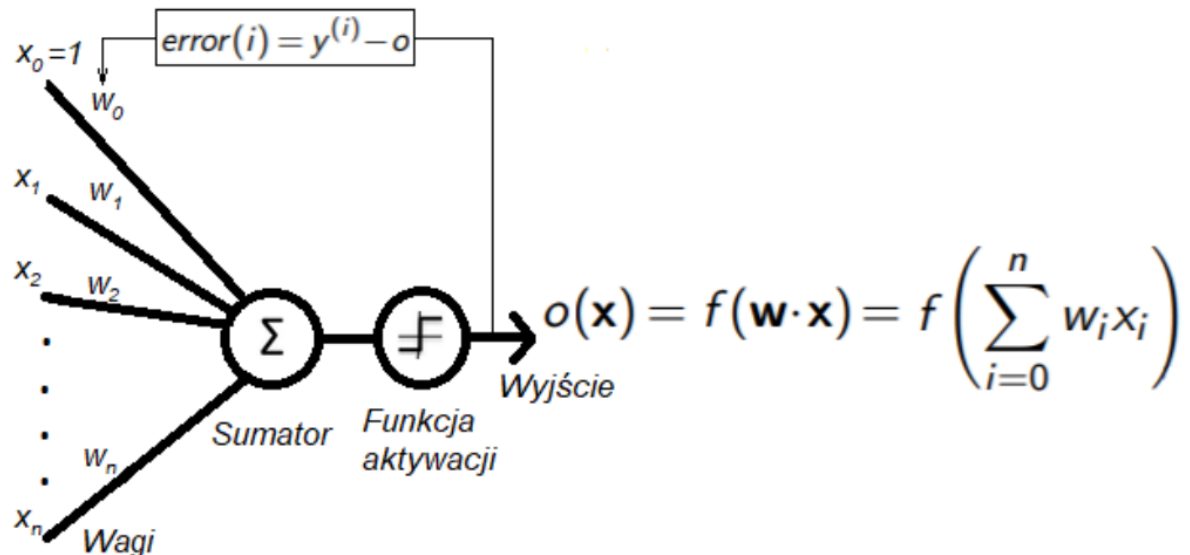


Perceptron oraz Adaline

1. Opis problemu Zadaniem była implementacja perceptronu i nauczenie go poprawnej klasyfikacji gatunków irysów, dla zbioru danych Iris.data. Neuron należało zaimplementować metodą perceptronu oraz Adaline, a następnie porównać klasyfikowanie na podstawie dwóch i trzech klas.

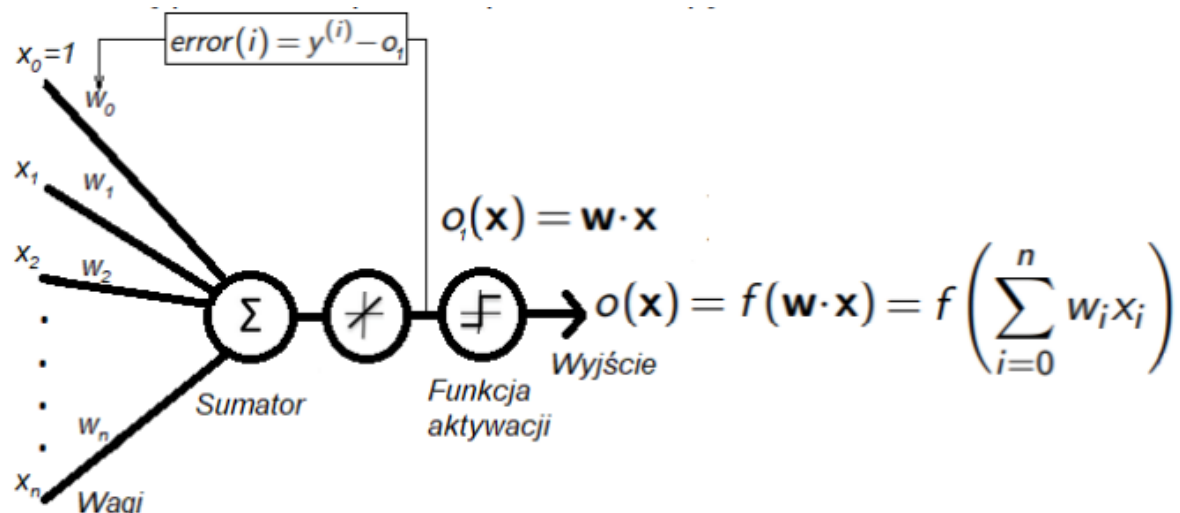
2. Przebieg zadania

1. Perceptron



Dla zbioru $Z = \{(x^{(1)}, y^{(1)}), \dots, (x^{(N)}, y^{(N)})\}$ od ustalonej liczby n_{epoch} należy iterować po zbiorze Z dla $i=1, \dots, N$ należy obliczyć wagi, $error(i) = y^{(i)} - o(x^{(i)})$ $\Delta w = \eta * error(i) * x^{(i)}$ $w = w + \Delta w$ η to współczynnik uczenia z przedziału $(0, 1)$. Zbiór Z to 80% danych z każdej klasy. Zbiór testowy to pozostałe 20%, na tym zbiorze testowany jest perceptron.

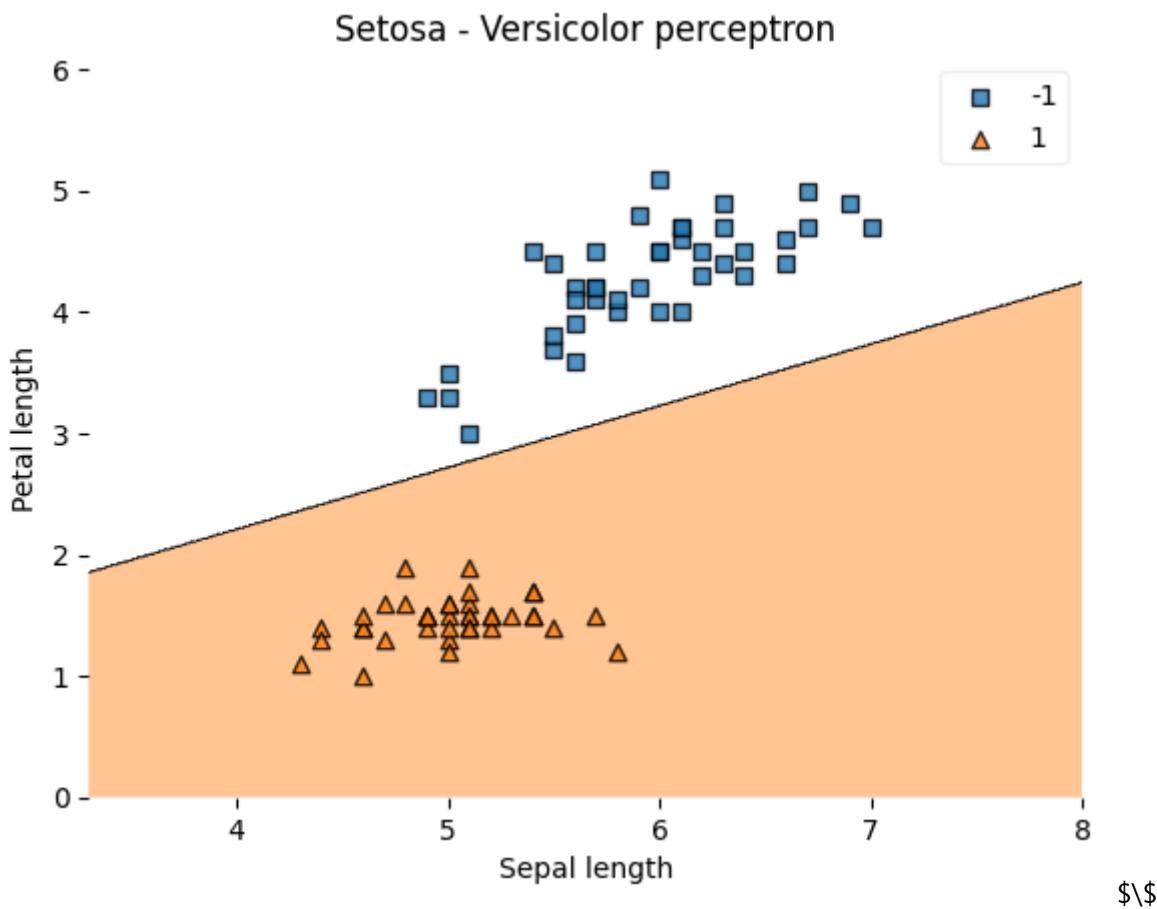
2. Adaline



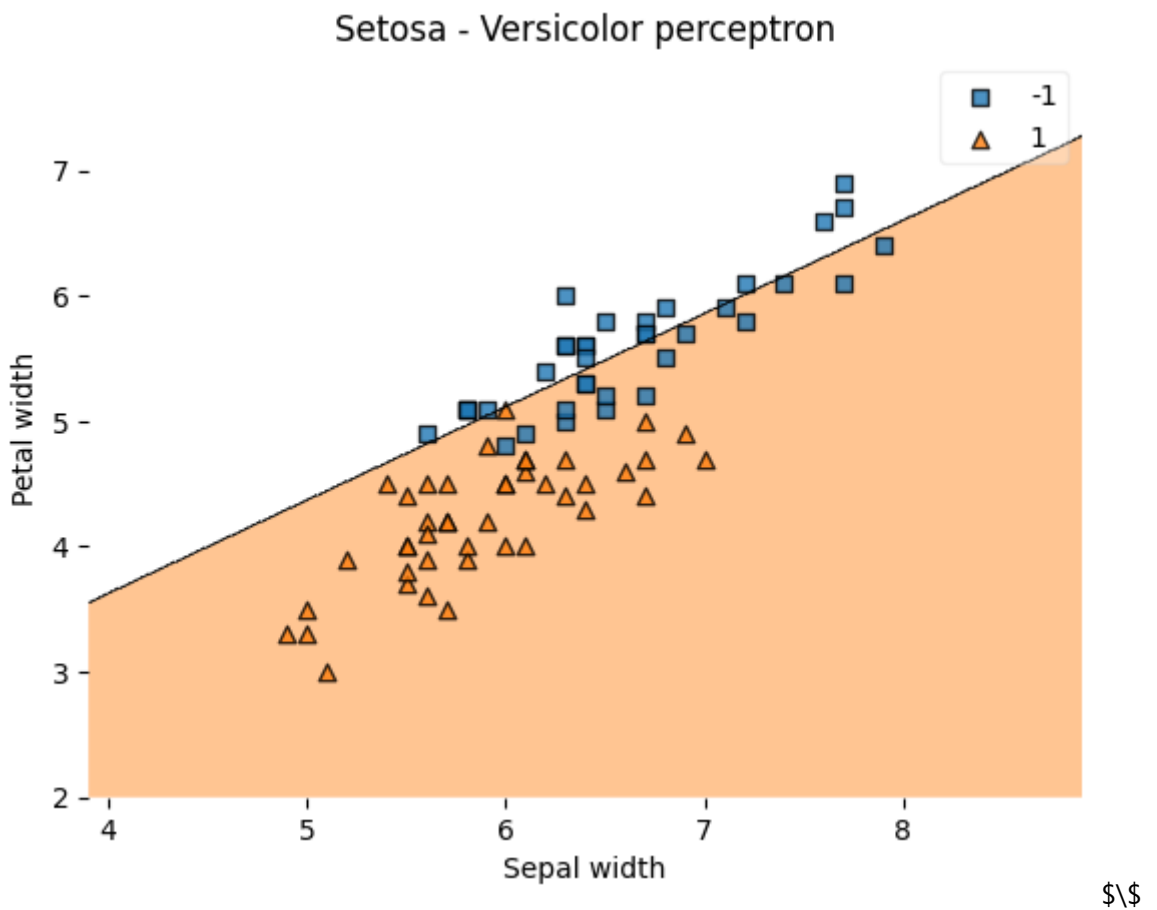
Dla zbioru $Z = \{(x^{(1)}, y^{(1)}), \dots, (x^{(N)}, y^{(N)})\}$ od ustalonej liczby n_{epoch} należy iterować po zbiorze Z dla $i=1, \dots, N$ należy obliczyć wagi, $error(i) = y^{(i)} - o_1(x^{(i)})$ $\Delta w = \eta * error(i) * x^{(i)}$ $w = w + \Delta w$ η to współczynnik uczenia z przedziału $(0, 1)$. Zbiór Z to 80% danych z każdej klasy. Zbiór testowy to pozostałe 20%, na tym zbiorze testowany jest Adaline.

3. Wyniki:

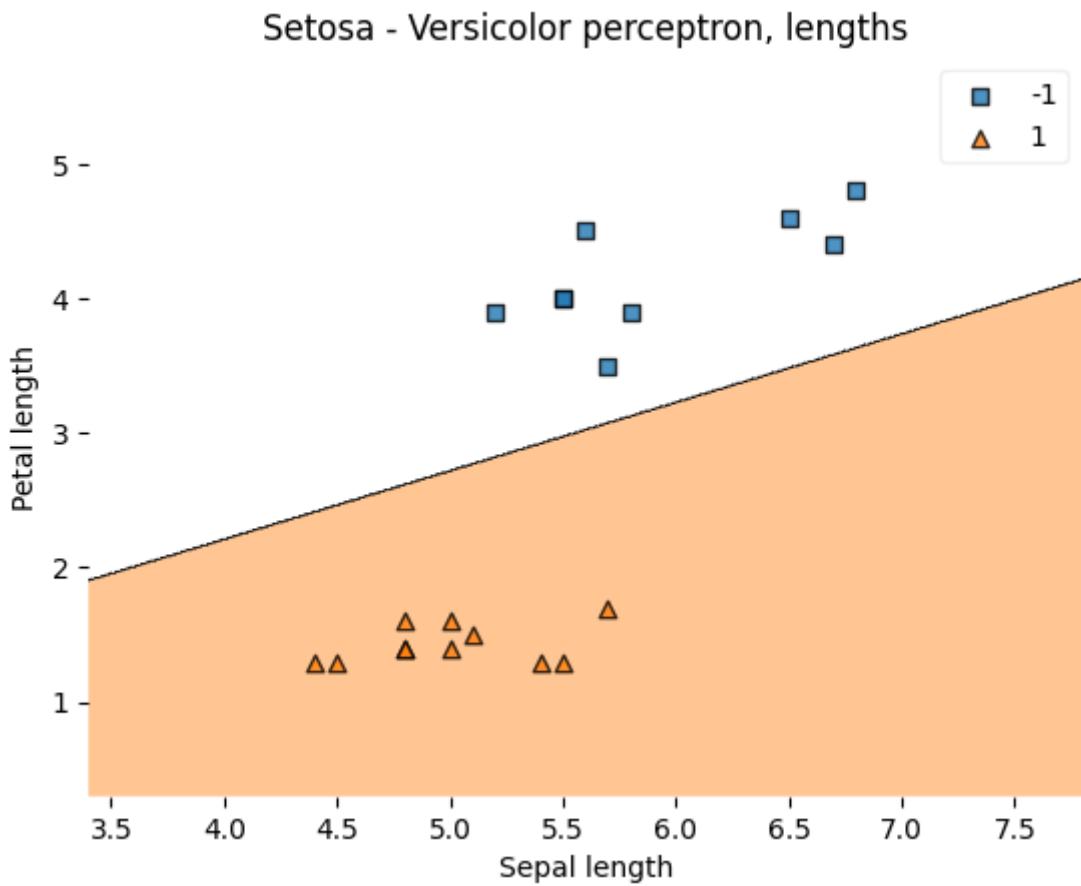
1. Perceptron, 2 klasy Trening na długościach:



Trening na szerokościach:

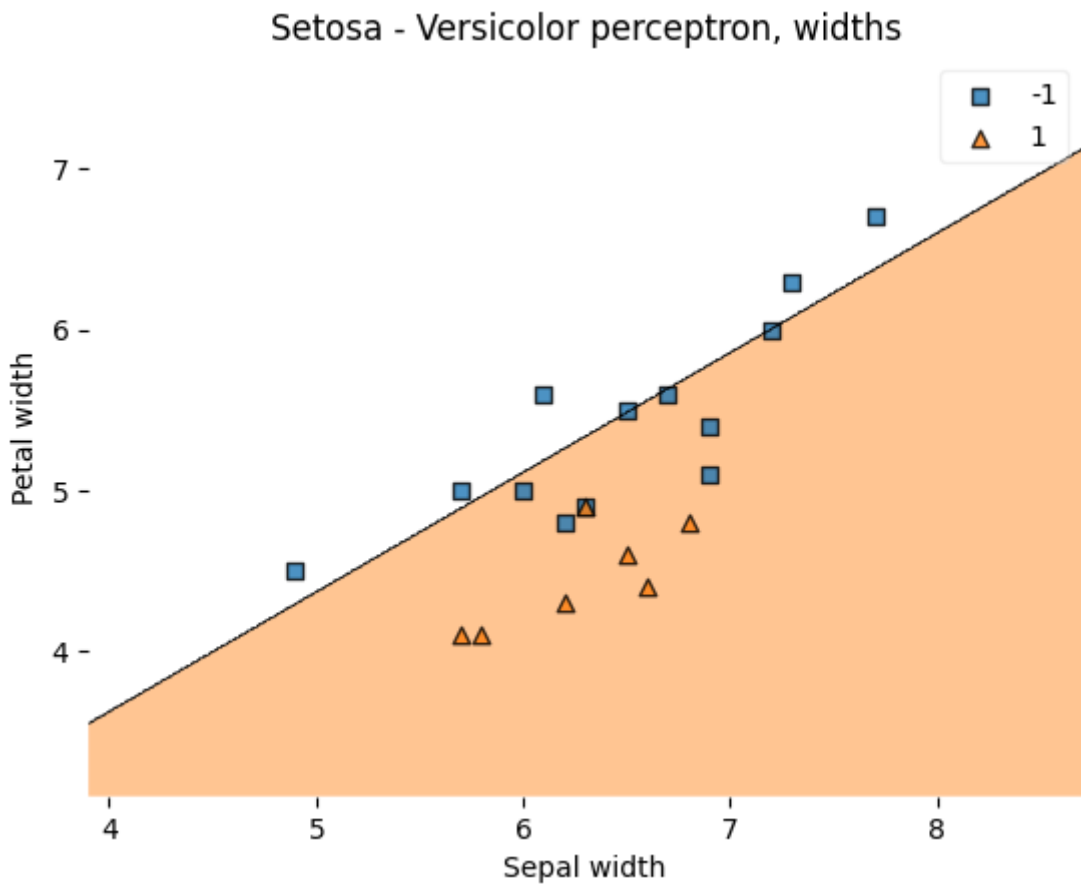


Test na długościach:



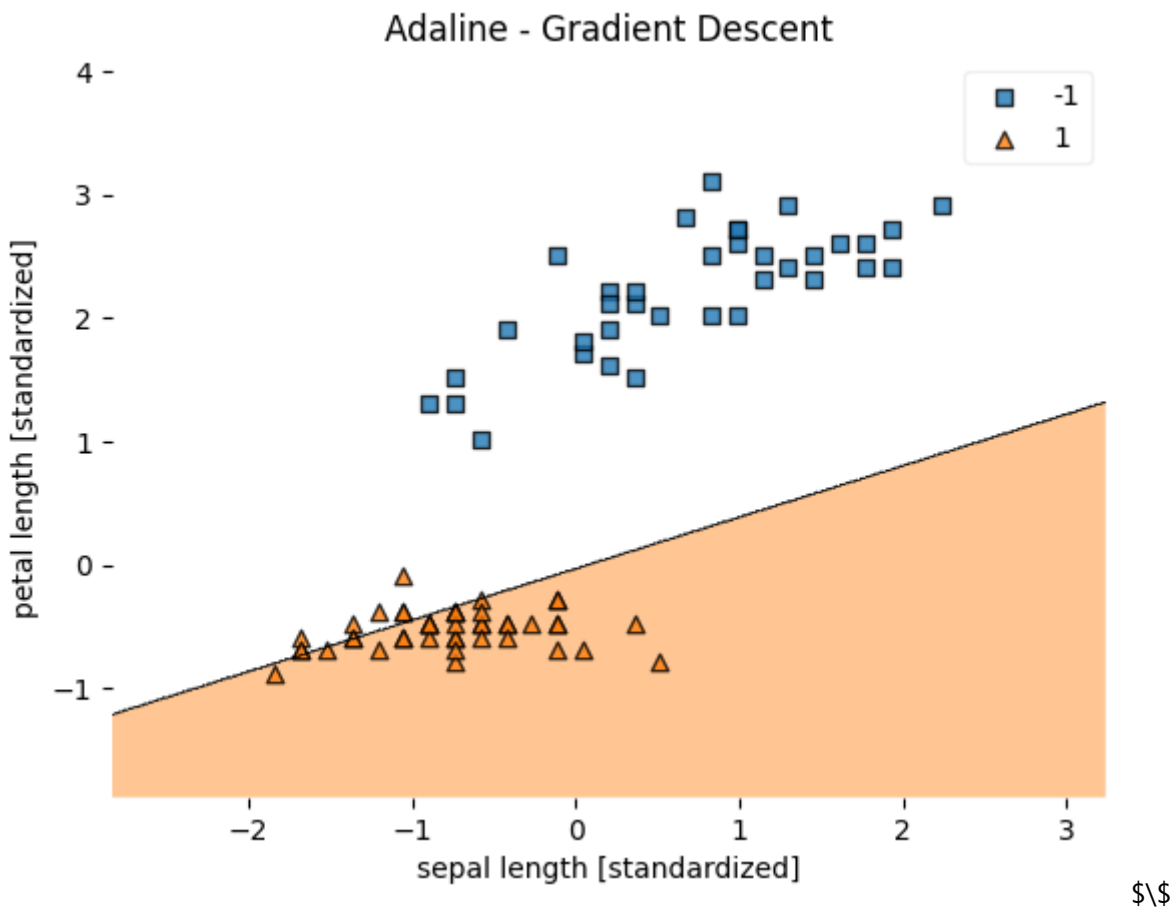
\$\$

Test na szerokościach: \$\$

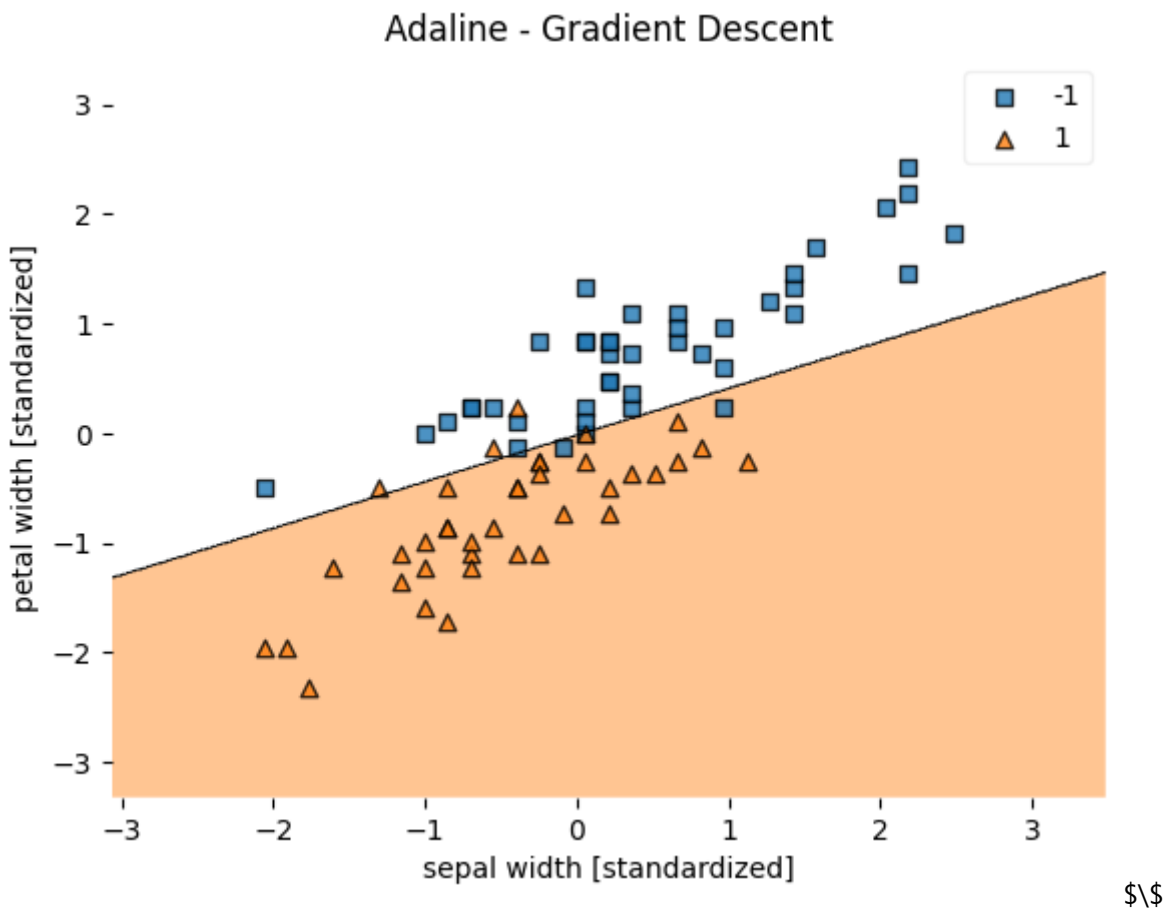


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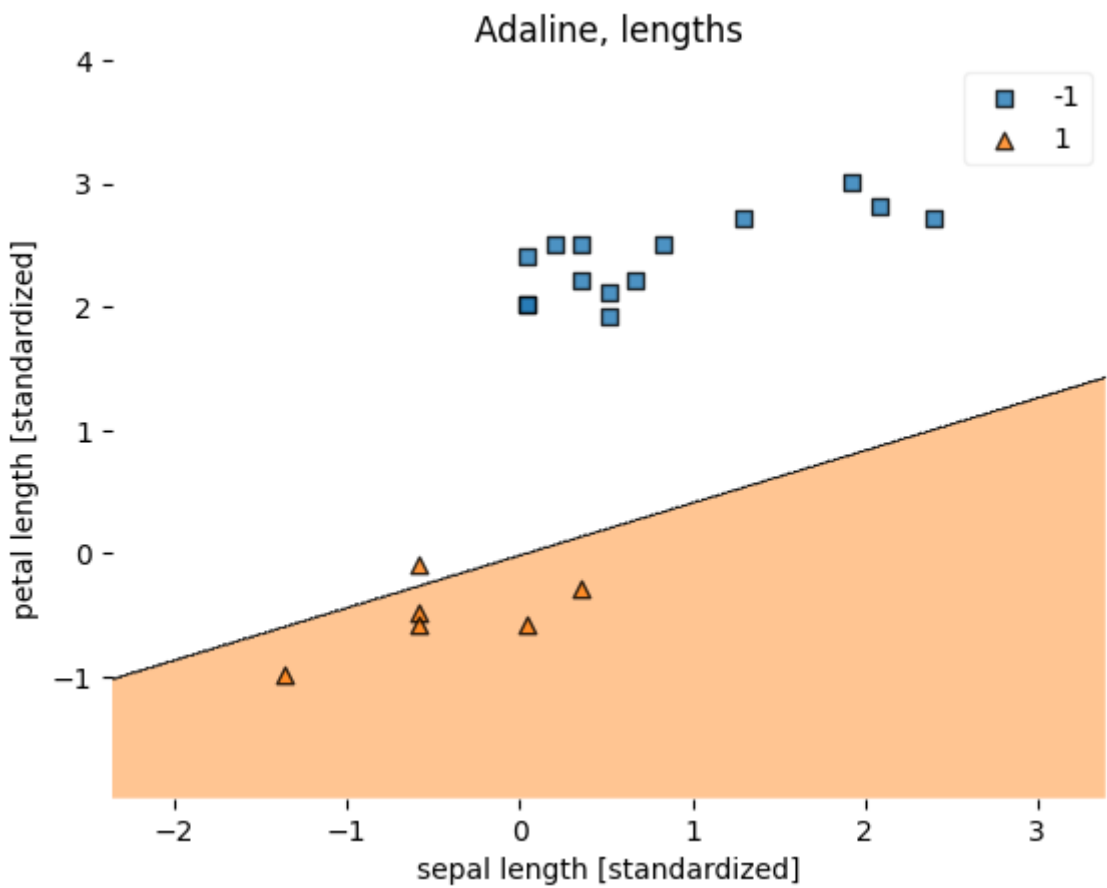
2. Adaline, 2 klasy $\$ \$$ Trening na długościach: $\$ \$$



Trening na szerokościach: $\$ \$$

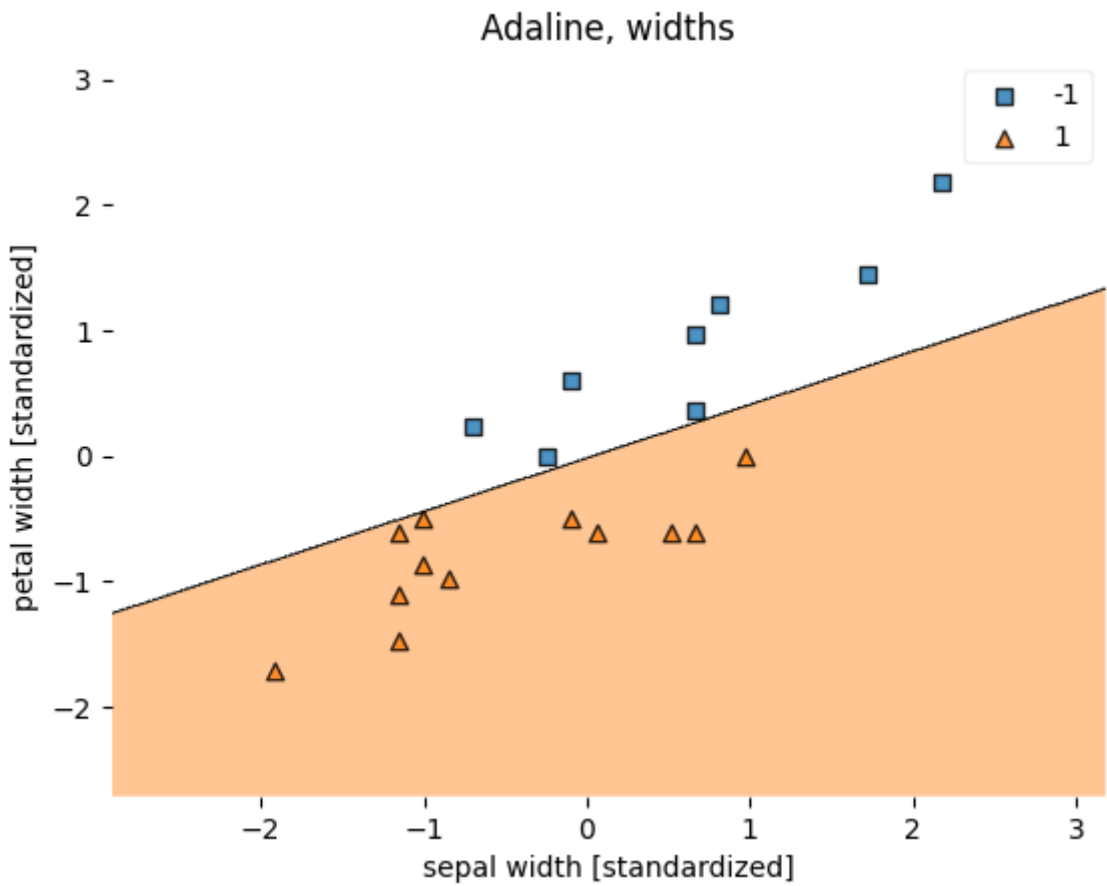


Test na długościach: $\$ \$$



\$\$

Test na szerokościach: \$\$



\$\$

3. Perceptron, 3 klasy \$\$\$ Liczba błędnych klasyfikacji: \$\$\$

```

1 ppn_seto = Perceptron(epochs=50, eta=0.001)
2 ppn_seto.train(X_seto_train, y_seto_train)
3
4 print(f"count of incorrect categorizations for 3 classes, lengths: {(y_seto_test != ppn_seto.predict(X_seto_test)).sum()} out of {len(y_seto_test)}")
[126] ✓ 0.0s
... count of incorrect categorizations for 3 classes, lengths: 0 out of 30

1 # dla klasy Virginica
2 ppn_virg = Perceptron(epochs=50, eta=0.001)
3 ppn_virg.train(X_virg_train, y_virg_train)
4
5 print(f"count of incorrect categorizations for 3 classes, lengths: {(y_virg_test != ppn_virg.predict(X_virg_test)).sum()} out of {len(y_virg_test)}")
[127] ✓ 0.1s
... count of incorrect categorizations for 3 classes, lengths: 2 out of 30

1 # dla klasy Versicolor
2 ppn_vers = Perceptron(epochs=50, eta=0.01)
3 ppn_vers.train(X_vers_train, y_vers_train)
4
5 print(f"count of incorrect categorizations for 3 classes, lengths: {(y_vers_test != ppn_vers.predict(X_vers_test)).sum()} out of {len(y_vers_test)}")
[128] ✓ 0.1s
... count of incorrect categorizations for 3 classes, lengths: 12 out of 30

```

\$\$\$

4. Adaline, 3 klasy \$\$\$ Liczba błędnych klasyfikacji: \$\$\$

```

1 #setosa
2 ada = AdalineGD (epochs=50, eta=0.01)
3 ada.train(X_seto_train, y_seto_train)
4
5 print(f"count of incorrect categorizations, adaline: {(y_seto_test != ada.predict(X_seto_test)).sum()} out of {len(y_seto_test)}")
[129] ✓ 0.0s
... count of incorrect categorizations, adaline: 21 out of 30

1 #virginica
2 ada = AdalineGD (epochs=50, eta=0.01)
3 ada.train(X_virg_train, y_virg_train)
4
5 print(f"count of incorrect categorizations, adaline: {(y_virg_test != ada.predict(X_virg_test)).sum()} out of {len(y_virg_test)}")
[130] ✓ 0.0s
... count of incorrect categorizations, adaline 21 out of 30

1 #versicolor
2 ada = AdalineGD (epochs=50, eta=0.01)
3 ada.train(X_vers_train, y_vers_train)
4
5 print(f"count of incorrect categorizations, adaline: {(y_vers_test != ada.predict(X_vers_test)).sum()} out of {len(y_vers_test)}")
[131] ✓ 0.0s
... count of incorrect categorizations, adaline 22 out of 30

```

\$\$\$

5. Różne kombinacje, perceptron. \$\$\$ Wpierw \$\$\$eta = 0.01\$, następnie \$\$\$eta = 0.5\$ \$\$\$ W kolejności setosa, virginica, versicolor \$\$\$ Liczba błędnych klasyfikacji: \$\$\$

```

Perceptron count of incorrect categorizations for Sepal length and sepal width: 1
Perceptron count of incorrect categorizations for petal length and petal width: 0
Perceptron count of incorrect categorizations for Sepal length and petal length: 0
Perceptron count of incorrect categorizations for Sepal length and petal width: 0
Perceptron count of incorrect categorizations for sepal width and petal length: 0
Perceptron count of incorrect categorizations for 4DIM: 0
Average accuracy Perceptron: 0.9944444444444445

Perceptron count of incorrect categorizations for Sepal length and sepal width: 15
Perceptron count of incorrect categorizations for petal length and petal width: 1
Perceptron count of incorrect categorizations for Sepal length and petal length: 1
Perceptron count of incorrect categorizations for Sepal length and petal width: 3
Perceptron count of incorrect categorizations for sepal width and petal length: 7
Perceptron count of incorrect categorizations for 4DIM: 4
Average accuracy Perceptron: 0.8277777777777778

Perceptron count of incorrect categorizations for Sepal length and sepal width: 17
Perceptron count of incorrect categorizations for petal length and petal width: 11
Perceptron count of incorrect categorizations for Sepal length and petal length: 9
Perceptron count of incorrect categorizations for Sepal length and petal width: 16
Perceptron count of incorrect categorizations for sepal width and petal length: 20
Perceptron count of incorrect categorizations for 4DIM: 12
Average accuracy Perceptron: 0.5277777777777778

Perceptron count of incorrect categorizations for Sepal length and sepal width: 0
Perceptron count of incorrect categorizations for petal length and petal width: 0
Perceptron count of incorrect categorizations for Sepal length and petal length: 0
Perceptron count of incorrect categorizations for Sepal length and petal width: 0
Perceptron count of incorrect categorizations for sepal width and petal length: 0
Perceptron count of incorrect categorizations for 4DIM: 0
Average accuracy Perceptron: 1.0

Perceptron count of incorrect categorizations for Sepal length and sepal width: 12
Perceptron count of incorrect categorizations for petal length and petal width: 6
Perceptron count of incorrect categorizations for Sepal length and petal length: 1
Perceptron count of incorrect categorizations for Sepal length and petal width: 2
Perceptron count of incorrect categorizations for sepal width and petal length: 1
Perceptron count of incorrect categorizations for 4DIM: 1
Average accuracy Perceptron: 0.8722222222222222

Perceptron count of incorrect categorizations for Sepal length and sepal width: 12
Perceptron count of incorrect categorizations for petal length and petal width: 8
Perceptron count of incorrect categorizations for Sepal length and petal length: 17
Perceptron count of incorrect categorizations for Sepal length and petal width: 10
Perceptron count of incorrect categorizations for sepal width and petal length: 9
Perceptron count of incorrect categorizations for 4DIM: 21
Average accuracy Perceptron: 0.5722222222222223

```

\$\$

6. Różne kombinacje, Adaline \$\$ Wpierw $\eta = 0.01$, następnie $\eta = 0.5$ \$\$ W kolejności setosa, virginica, versicolor \$\$ Liczba błędnych klasyfikacji: \$\$

```

Adaline count of incorrect categorizations for Sepal length and sepal width: 0
Adaline count of incorrect categorizations for petal length and petal width: 27
Adaline count of incorrect categorizations for Sepal length and petal length: 28
Adaline count of incorrect categorizations for Sepal length and petal width: 27
Adaline count of incorrect categorizations for sepal width and petal length: 0
Adaline count of incorrect categorizations for 4DIM: 18
Average accuracy Adaline: 0.4444444444444444

Adaline count of incorrect categorizations for Sepal length and sepal width: 4
Adaline count of incorrect categorizations for petal length and petal width: 24
Adaline count of incorrect categorizations for Sepal length and petal length: 21
Adaline count of incorrect categorizations for Sepal length and petal width: 26
Adaline count of incorrect categorizations for sepal width and petal length: 2
Adaline count of incorrect categorizations for 4DIM: 17
Average accuracy Adaline: 0.4777777777777777

Adaline count of incorrect categorizations for Sepal length and sepal width: 7
Adaline count of incorrect categorizations for petal length and petal width: 17
Adaline count of incorrect categorizations for Sepal length and petal length: 20
Adaline count of incorrect categorizations for Sepal length and petal width: 16
Adaline count of incorrect categorizations for sepal width and petal length: 5
Adaline count of incorrect categorizations for 4DIM: 21
Average accuracy Adaline: 0.5222222222222223

Adaline count of incorrect categorizations for Sepal length and sepal width: 27
Adaline count of incorrect categorizations for petal length and petal width: 29
Adaline count of incorrect categorizations for Sepal length and petal length: 26
Adaline count of incorrect categorizations for Sepal length and petal width: 23
Adaline count of incorrect categorizations for sepal width and petal length: 28
Adaline count of incorrect categorizations for 4DIM: 23
Average accuracy Adaline: 0.1333333333333333

Adaline count of incorrect categorizations for Sepal length and sepal width: 22
Adaline count of incorrect categorizations for petal length and petal width: 20
Adaline count of incorrect categorizations for Sepal length and petal length: 26
Adaline count of incorrect categorizations for Sepal length and petal width: 23
Adaline count of incorrect categorizations for sepal width and petal length: 23
Adaline count of incorrect categorizations for 4DIM: 20
Average accuracy Adaline: 0.25555555555555554

Adaline count of incorrect categorizations for Sepal length and sepal width: 17
Adaline count of incorrect categorizations for petal length and petal width: 17
Adaline count of incorrect categorizations for Sepal length and petal length: 18
Adaline count of incorrect categorizations for Sepal length and petal width: 11
Adaline count of incorrect categorizations for sepal width and petal length: 21
Adaline count of incorrect categorizations for 4DIM: 19
Average accuracy Adaline: 0.4277777777777778

```

\$\backslash\$

4. Wnioski:

- współczynnik uczenia nie ma większego wpływu na klasyfikację trzech klas metodą perceptronu.
- Adaline sprawdza się lepiej przy dwóch klasach
- Perceptron sprawdza się lepiej przy trzech klasach