```
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
sns.set style('darkgrid')
np.random.seed(42)
# Generowanie danych
nb of samples = 30
sequence len = 20
# Możliwe wartości
levels = np.array([0.33, 0.66, 1.0])
# Tworzymy dane wejściowe (X)
X = np.random.choice(levels, size=(nb of samples, sequence len))
# Cele: średnie odchylenie standardowe
t = np.std(X, axis=1)
# Forward propagation
def update state(xk, sk, wx, wRec):
    return xk * wx + sk * wRec
def forward states(X, wx, wRec):
    S = np.zeros((X.shape[0], X.shape[1]+1))
    for k in range(X.shape[1]):
        S[:,k+1] = update state(X[:,k], S[:,k], wx, wRec)
    return S
def loss(y, t):
    return np.mean((t - y)**2)
# Backward propagation
def output gradient(y, t):
    return 2. * (y - t)
def backward gradient(X, S, grad out, wRec):
    grad over time = np.zeros((X.shape[0], X.shape[1]+1))
    grad over time[:,-1] = grad out
    wx grad = 0
    wRec_grad = 0
    for k in range(X.shape[1], 0, -1):
        wx grad += np.sum(np.mean(grad over time[:,k] * X[:,k-1],
axis=0)
        wRec grad += np.sum(np.mean(grad over time[:,k] * S[:,k-1],
axis=0))
        grad over time[:,k-1] = grad over time[:,k] * wRec
    return (wx grad, wRec grad), grad over time
```

```
# Sprawdzenie gradientu
params = [1.0, 1.0]
eps = 1e-7
S = forward states(X, params[0], params[1])
grad out = output gradient(S[:,-1], t)
backprop_grads, _ = backward_gradient(X, S, grad_out, params[1])
for p idx in range(len(params)):
    original = params[p idx]
    params[p idx] += eps
    plus loss = loss(forward states(X, params[0], params[1])[:,-1], t)
    params[p idx] -= 2 * eps
    min loss = loss(forward states(X, params[\frac{0}{0}], params[\frac{1}{1}])[:,-\frac{1}{1}, t)
    params[p idx] = original
    grad_num = (plus_loss - min_loss) / (2 * eps)
    if not np.isclose(grad num, backprop grads[p idx]):
        raise ValueError(f'Gradient check failed at param {p idx}')
print("Gradient check passed!")
Gradient check passed!
# Optymalizacja RProp
def update_rprop(X, t, W, W_prev_sign, W_delta, eta_p, eta_n):
    S = forward states(X, W[0], W[1])
    grad out = \overline{\text{output gradient}}(S[:,-1], t)
    W grads, = backward gradient(X, S, grad out, W[1])
    W sign = np.sign(W grads)
    for i in range(len(W)):
        if W sign[i] == W prev sign[i]:
            W delta[i] *= eta p
        else:
            W delta[i] *= eta n
    return W delta, W sign
# Parametry RProp
eta p = 1.2
eta n = 0.5
W = [0.5, 0.5]
W \text{ delta} = [0.001, 0.001]
W \text{ sign} = [0, 0]
ws history = [tuple(W)]
for in range(300):
    W delta, W sign = update rprop(X, t, W, W sign, W delta, eta p,
eta n)
    for i in range(len(W)):
        W[i] -= W sign[i] * W delta[i]
    ws history.append(tuple(W))
```

```
print(f"[] Final weights: wx = {W[0]:.4f}, wRec = {W[1]:.4f}")

[] Final weights: wx = 0.0473, wRec = 0.8967

# Testowanie na nowej sekwencji
test_input = np.random.choice(levels, size=(1, 20))
expected_output = np.std(test_input)
model_output = forward_states(test_input, W[0], W[1])[:,-1][0]

print(f"Target (std): {expected_output:.4f}")
print(f"Model output: {model_output:.4f}")

Target (std): 0.2331
Model output: 0.2433
```