

UPN Intelligent Systems Course

Final Grade Project 2020

Premise

Train a NN to determine if student will approve a course, based on this rules:

- Consider three test results.
- The student will have almost two approved units to pass the course.
- To get the final result, there's no math calculation. Just take the mentioned previously.
- Results should be between 0 to 20.

► TaskLocalRNG()

```
• begin
•   using Flux ✓
•   using Plots ✓
•   using Random ✓
•   using Metrics ✓
•   using MLUtils ✓
•   using Statistics ✓
•   using ProgressLogging ✓
•   Random.seed!(9)
• end
```

Generating grades data

generatedata (generic function with 1 method)

```
• generatedata(::Int64) = rand(0:20, 3) |> Vector{Int32}
```

checkcriteria (generic function with 1 method)

```
• checkcriteria(t) = (map(x -> x>11, t) |> sum > 1) |> Int32
```

n_student = 100000

```
• n_student = 100_000
```

generated_data =

```
► [[15, 4, 16], [10, 14, 7], [18, 7, 10], [10, 3, 12], [16, 4, 20], [20, 8, 3], [12, 0, 9]]
• generated_data = map(generatedata, 1:n_student)
```

```
data_x =
3×100000 Matrix{Int32}:
 15  10  18  10  16  20  12  17  20  14  ...  7  18  5  5  16  15  1  15  5  11  14
  4  14  7  3  4  8  0  3  1  0  ...  0  3  1  2  4  13  17  8  3  15  6
 16  7  10  12  20  3  9  18  8  6  ...  7  15  13  2  0  1  15  12  5  0  9
```

- `data_x = reduce(hcat, generated_data)`

```
data_y =
1×100000 Matrix{Int32}:
 1  0  0  0  1  0  0  1  0  0  1  0  0  0  ...  0  0  0  1  0  0  0  1  1  1  0  0  0
```

- `data_y = generated_data' .|> checkcriteria`

Preparing data

```
norm_data_x =
3×100000 Matrix{Float32}:
 0.827859  0.00169199  1.32356  0.00169199  ... -0.824475  0.166925  0.662625
-0.988051  0.664125  -0.492398  -1.15327  ... -1.15327  0.829343  -0.657616
 0.988098  -0.496873  -0.00188261  0.328111  ... -0.826867  -1.65185  -0.166879
```

- `norm_data_x = Flux.normalise(data_x) |> Matrix{Float32}`

```
► ((3×80000 view(::Matrix{Float32}, :, 1:80000) with eltype Float32: , 1×80000
 0.827859  0.00169199  1.32356  ... -0.824475  -0.163541  0.827859  1  0  (
-0.988051  0.664125  -0.492398  0.00325479  -1.15327  1.65543
 0.988098  -0.496873  -0.00188261  1.31809  -1.48685  1.15309
```

- `train_data, test_data = MLUtils.splitobs((norm_data_x, data_y); at=0.8)`

```
train_data_loader =
80-element DataLoader(::Tuple{SubArray{Float32, 2, Matrix{Float32}}, Tuple{Base.Slice{B
with first element:
(3×1000 Matrix{Float32}, 1×1000 Matrix{Int32},)
```

- `train_data_loader = DataLoader(train_data, batchsize=1000)`

```
test_data_loader =
20000-element DataLoader(::Tuple{SubArray{Float32, 2, Matrix{Float32}}, Tuple{Base.Slic
with first element:
(3×1 Matrix{Float32}, 1×1 Matrix{Int32},)
```

- `test_data_loader = DataLoader(test_data)`

Defining model

```
► Adam(0.05, (0.9, 0.999), 1.0e-8, IdDict())
```

- `begin`
- `model = Chain(`
- `Dense(3 => 8, relu),`
- `Dense(8 => 1, sigmoid)`
- `) |> gpu`
- `optimizer = Adam(0.05)`
- `end`

```
Chain(
  Dense(3 => 8, relu),          # 32 parameters
  Dense(8 => 1,  $\sigma$ ),        # 9 parameters
)                               # Total: 4 arrays, 41 parameters, 420 bytes.
```

- `model`

loss (generic function with 1 method)

- `loss(x, y) = Flux.binarycrossentropy(model(x), y) |> gpu`

Training phase

```
epochs = 15
```

- `epochs = 15`

```
• begin
•   train_losses = Float32[]
•   test_losses = Float32[]
•
•   @progress for e in 1:epochs
•     for d in train_data_loader
•       gs = gradient(Flux.params(model)) do
•         l = loss(d...)
•       end
•       Flux.update!(optimizer, Flux.params(model), gs)
•     end
•     push!(train_losses, loss(train_data_loader.data...))
•     push!(test_losses, loss(test_data_loader.data...))
•   end
• end
```

100%

Testing phase

accuracy (generic function with 1 method)

- `accuracy(y, Y) = mean(Y .== y)`

```
• begin
•   test_pred = Int32[]
•   test_truth = test_data_loader.data[2]
•
•   for (x, y) in test_data_loader
•     pred = round(Int32, model(x)[1])
•     append!(test_pred, pred)
•   end
• end
```

```
incorrect_predictions = 0
```

- `incorrect_predictions = test_truth .- test_pred' |> sum |> abs`

```
• @info "Correct predictions: $(size(test_pred)[1] - incorrect_predictions) -  
  Incorrect: $incorrect_predictions"
```

```
Correct predictions: 20000 - Incorrect: 0
```

```
mae = 0.0
```

```
• mae = Metrics.mae(test_pred', test_truth)
```

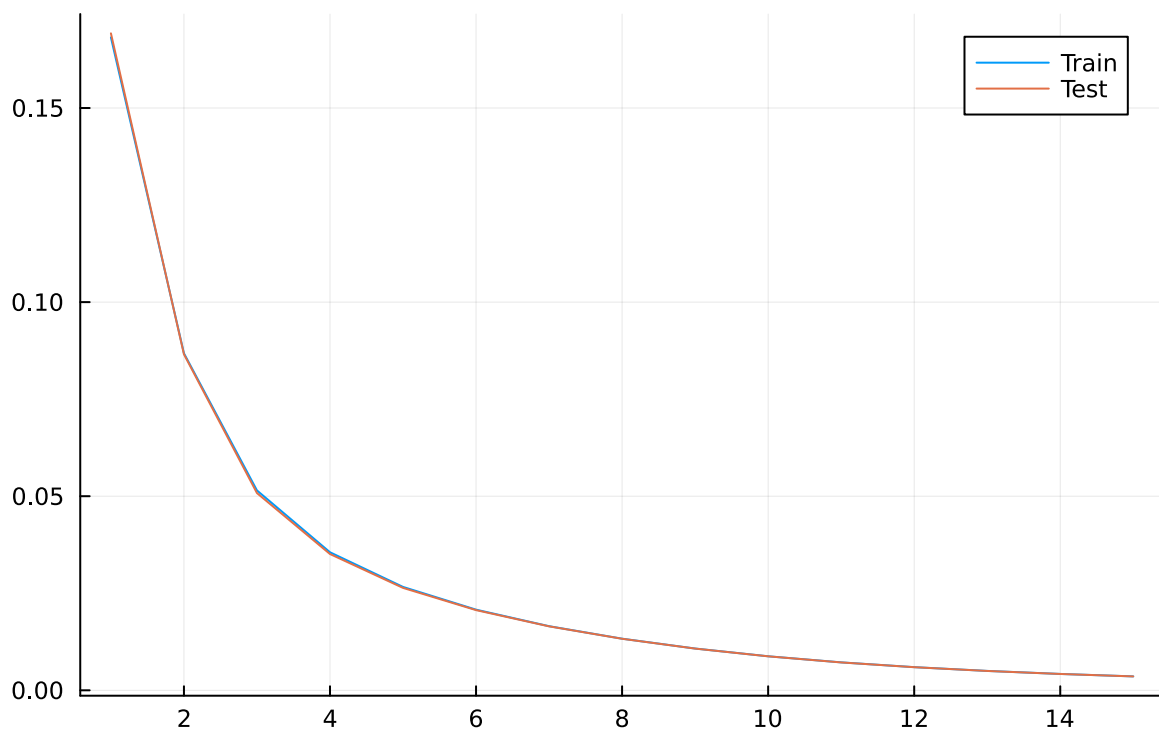
```
mse = 0.0
```

```
• mse = Metrics.mse(test_pred', test_truth)
```

```
acc = 1.0
```

```
• acc = accuracy(test_pred', test_truth)
```

Loss over epochs: 15



```
• begin  
• plot(1:epochs, train_losses, title="Loss over epochs: $epochs", label="Train")  
• plot!(1:epochs, test_losses, label="Test")  
• end
```