Machine Translation Sabrina Brändle, Eirini Valkana

**MT Exercise 3  
Recurrent networks**

**Link to ‘ex3’ repository:** [**https://github.com/sabrinabraendle/mt-exercise-3**](https://github.com/sabrinabraendle/mt-exercise-3)

**Training perplexity after each epoch**

|  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- |
| Train. perplexity | dropout 0 | dropout 0.3 | dropout 0.5 | dropout 0.7 | dropout 1 |
| Epoch 1 | 251.43 | 260.83 | 275.33 | 308.2 | 464.75 |
| Epoch 2 | 114.1 | 129.76 | 148.67 | 189.31 | 428.07 |
| Epoch 3 | 83.55 | 99.61 | 116.31 | 158.08 | 418.49 |
| Epoch 4 | 68.02 | 84.45 | 100.64 | 140.97 | 414.37 |
| Epoch 5 | 57.38 | 74.67 | 91.06 | 131.23 | 412.15 |
| Epoch 6 | 49.04 | 67.75 | 85.02 | 123.69 | 410.8 |
| Epoch 7 | 41.77 | 62.18 | 79.72 | 118.65 | 409.91 |
| Epoch 8 | 35.51 | 57.64 | 76.53 | 114.47 | 409.29 |
| Epoch 9 | 30.42 | 54.07 | 72.94 | 111.49 | 408.85 |
| Epoch 10 | 26.46 | 51.49 | 70.38 | 108.48 | 405.55 |
| Epoch 11 | 23.28 | 48.78 | 68.72 | 106.8 | 405.55 |
| Epoch 12 | 20.75 | 46.59 | 66.66 | 104.77 | 405.55 |
| Epoch 13 | 18.77 | 44.59 | 64.94 | 103.48 | 405.54 |
| Epoch 14 | 17.23 | 43.04 | 63.7 | 100.96 | 405.51 |
| Epoch 15 | 16.03 | 41.52 | 62.8 | 100.62 | 405.48 |
| Epoch 16 | 14.95 | 40.18 | 61.23 | 99.33 | 404.71 |
| Epoch 17 | 14.09 | 39.06 | 60.52 | 98.24 | 404.73 |
| Epoch 18 | 13.38 | 37.81 | 59.5 | 97.71 | 404.45 |
| Epoch 19 | 12.7 | 37.11 | 58.33 | 96.61 | 404.47 |
| Epoch 20 | 12.07 | 36.14 | 57.64 | 96.17 | 404.49 |
| Epoch 21 | 11.67 | 35.45 | 56.69 | 95.8 | 404.5 |
| Epoch 22 | 11.23 | 34.81 | 56.57 | 95.14 | 404.51 |
| Epoch 23 | 10.89 | 34.04 | 55.57 | 94.69 | 404.41 |
| Epoch 24 | 10.51 | 33.5 | 54.68 | 93.99 | 404.41 |
| Epoch 25 | 10.31 | 32.89 | 54.53 | 93.48 | 404.42 |
| Epoch 26 | 9.95 | 32.36 | 54.09 | 92.78 | 404.42 |
| Epoch 27 | 9.66 | 31.82 | 53.74 | 91.81 | 404.43 |
| Epoch 28 | 9.48 | 31.4 | 53.02 | 91.19 | 404.43 |
| Epoch 29 | 9.23 | 31.11 | 52.5 | 91.27 | 404.43 |
| Epoch 30 | 9.03 | 30.76 | 52.4 | 90.89 | 404.44 |
| Epoch 31 | 8.89 | 30.53 | 51.66 | 89.88 | 404.44 |
| Epoch 32 | 8.63 | 29.99 | 51 | 90.8 | 404.44 |
| Epoch 33 | 8.56 | 29.85 | 51.25 | 89.76 | 404.44 |
| Epoch 34 | 8.45 | 24.92 | 51.05 | 89.26 | 404.44 |
| Epoch 35 | 8.25 | 24.12 | 50.52 | 89.2 | 404.45 |
| Epoch 36 | 8.04 | 23.63 | 50.14 | 82.03 | 404.45 |
| Epoch 37 | 8.01 | 23.04 | 49.75 | 80.32 | 404.45 |
| Epoch 38 | 7.89 | 22.8 | 49.44 | 78.78 | 404.45 |
| Epoch 39 | 7.75 | 22.58 | 49.27 | 78.06 | 404.42 |
| Epoch 40 | 5.28 | 22.36 | 49.2 | 77.19 | 404.42 |

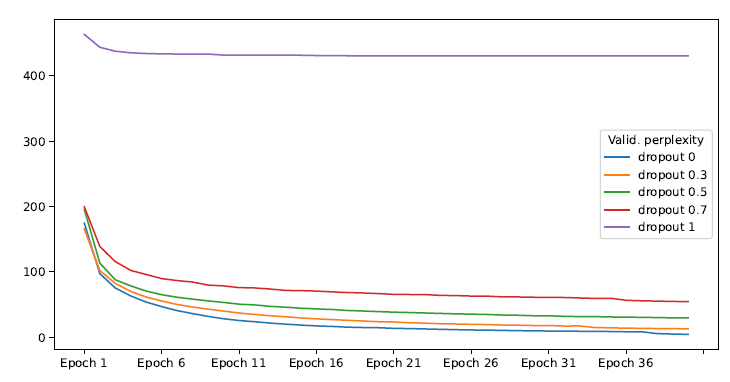
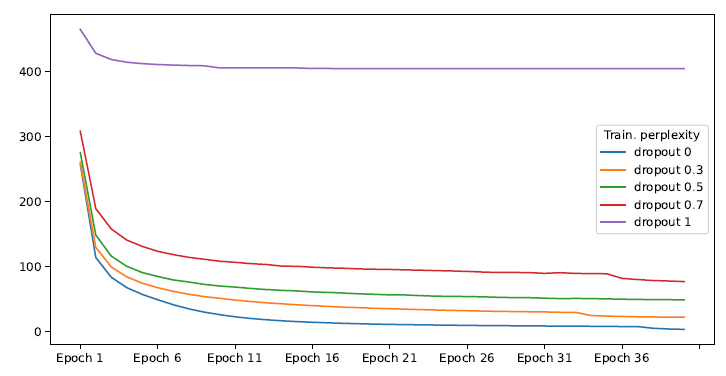
**Validation perplexity on the validation set after each epoch**

|  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- |
| Valid. perplexity | dropout 0 | dropout 0.3 | dropout 0.5 | dropout 0.7 | dropout 1 |
| Epoch 1 | 186.16 | 165.24 | 195.47 | 199.31 | 463.2 |
| Epoch 2 | 96.17 | 101.9 | 113.23 | 138.69 | 443.49 |
| Epoch 3 | 78.86 | 82.12 | 87.6 | 115.48 | 437.53 |
| Epoch 4 | 65.72 | 69.79 | 78.29 | 101.9 | 435.07 |
| Epoch 5 | 53.86 | 61.21 | 70.35 | 95.68 | 433.94 |
| Epoch 6 | 46.37 | 55.14 | 64.74 | 89.39 | 433.4 |
| Epoch 7 | 40.28 | 49.76 | 60.86 | 86.39 | 433.17 |
| Epoch 8 | 34.59 | 45.93 | 57.92 | 84.12 | 433.1 |
| Epoch 9 | 30.43 | 42.63 | 55.27 | 79.51 | 433.13 |
| Epoch 10 | 27.74 | 39.69 | 52.91 | 78.25 | 431.48 |
| Epoch 11 | 24.92 | 36.72 | 50.26 | 75.59 | 431.4 |
| Epoch 12 | 22.58 | 34.54 | 49.11 | 75.11 | 431.36 |
| Epoch 13 | 21.06 | 32.56 | 46.95 | 73.48 | 431.34 |
| Epoch 14 | 19.66 | 30.99 | 45.67 | 71.33 | 431.33 |
| Epoch 15 | 18.11 | 29.15 | 44.1 | 70.97 | 431.33 |
| Epoch 16 | 16.77 | 27.81 | 43.01 | 70.32 | 430.86 |
| Epoch 17 | 16.07 | 26.67 | 42.15 | 69.07 | 430.86 |
| Epoch 18 | 15.24 | 25.56 | 40.63 | 68.07 | 430.52 |
| Epoch 19 | 14.38 | 24.51 | 39.81 | 67.41 | 430.43 |
| Epoch 20 | 13.62 | 23.37 | 38.86 | 66.44 | 430.4 |
| Epoch 21 | 12.85 | 22.98 | 38.07 | 65.32 | 430.4 |
| Epoch 22 | 12.14 | 21.78 | 37.46 | 65.08 | 430.4 |
| Epoch 23 | 11.6 | 21.29 | 36.79 | 64.88 | 430.37 |
| Epoch 24 | 11.17 | 20.27 | 36.09 | 63.78 | 430.34 |
| Epoch 25 | 10.87 | 19.92 | 35.44 | 63.36 | 430.32 |
| Epoch 26 | 10.68 | 19.49 | 35.02 | 62.69 | 430.3 |
| Epoch 27 | 10.25 | 19.02 | 34.48 | 62.65 | 430.29 |
| Epoch 28 | 9.93 | 18.36 | 33.65 | 61.35 | 430.28 |
| Epoch 29 | 9.58 | 18.09 | 33.33 | 61.31 | 430.28 |
| Epoch 30 | 9.31 | 17.6 | 32.54 | 60.9 | 430.27 |
| Epoch 31 | 9.13 | 17.47 | 32.42 | 60.56 | 430.27 |
| Epoch 32 | 8.78 | 16.84 | 31.68 | 60.49 | 430.27 |
| Epoch 33 | 8.64 | 16.9 | 31.22 | 59.75 | 430.27 |
| Epoch 34 | 8.41 | 14.53 | 31.22 | 59.1 | 430.26 |
| Epoch 35 | 8.24 | 14 | 30.69 | 59.34 | 430.26 |
| Epoch 36 | 8.09 | 13.56 | 30.37 | 56.2 | 430.26 |
| Epoch 37 | 7.81 | 13.3 | 30.2 | 55.41 | 430.26 |
| Epoch 38 | 7.69 | 12.98 | 29.72 | 55 | 430.26 |
| Epoch 39 | 7.71 | 12.74 | 29.33 | 54.49 | 430.26 |
| Epoch 40 | 5.03 | 12.52 | 29.21 | 54.04 | 430.26 |

**Test perplexity on the test set as the very last step in training**

|  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- |
| Test perplexity | dropout 0 | dropout 0.3 | dropout 0.5 | dropout 0.7 | dropout 1 |
| Epoch 40 | 5.29 | 13.37 | 30.62 | 55.39 | 415.02 |

**Create a line chart each for the training and the validation perplexity to visualize the results.**

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**Can you see a connection between the training, validation, and test perplexity? Based on your results, which dropout setting do you think is the best and why?**

In all sets it is observed that as the dropout rate increases the perplexity value increases as well. Dropout rate 0.0 gives the lowest test perplexity score, so given our model and dataset it works the best. If the neural network was deeper (more than 2 layers, which is the default, and more than 200 units in the hidden layer) maybe a higher dropout rate would yield better results.

**Sample some text from your best model (the one that obtains the lowest test perplexity), for instance by changing the script scripts/generate.sh. What do you think of its quality? Does it resemble the original training data?**

The vocabulary is matched quite well, and some parts of the generated text make sense syntactically. Nevertheless, the sentences as a whole do not have meaning yet. Therefore, despite the fact that the vocabulary and style resemble the training data, the quality is not entirely met with this model.

**Link to ‘examples’ repository:** [**https://github.com/sabrinabraendle/examples**](https://github.com/sabrinabraendle/examples)