

## Contents

<b>1</b>	<b>Loss Interpretation Quick Reference Card</b>	<b>1</b>
1.1	Loss Value Interpretation . . . . .	1
1.2	Quick Examples . . . . .	2
1.2.1	Confident & Correct (Best) . . . . .	2
1.2.2	Uncertain & Correct (OK) . . . . .	2
1.2.3	Confident & Wrong (Bad) . . . . .	2
1.3	Training Health Checklist . . . . .	2
1.3.1	✓ Healthy Training . . . . .	2
1.3.2	⚠ Warning Signs . . . . .	2
1.4	Target Metrics (Your Model) . . . . .	3
1.5	When to Stop Training . . . . .	3
1.5.1	Stop Now ☐ . . . . .	3
1.5.2	Keep Training ☐ . . . . .	3
1.6	Loss vs Accuracy Decision Matrix . . . . .	3
1.7	What Loss Tells You . . . . .	4
1.8	Common Questions . . . . .	4
1.9	At a Glance . . . . .	5

## 1 Loss Interpretation Quick Reference Card

**For:** SKOL RNN 3-Class Text Classification (Nomenclature, Description, Misc)

### 1.1 Loss Value Interpretation

Loss	Meaning	Confidence	Action
<b>0.0 - 0.1</b>	Perfect	Extremely confident	⚠ Check for overfitting
<b>0.1 - 0.4</b>	Excellent	Very confident	✓ Great model!
<b>0.4 - 0.7</b>	Good	Reasonably confident	✓ Good for production
<b>0.7 - 1.1</b>	Fair	Uncertain	☐ Train more
<b>1.1 - 1.5</b>	Poor	Close to random	⚠ Check model/data

Loss	Meaning	Confidence	Action
<b>1.5+</b>	Very Poor	Confidently wrong	<b>✗</b> Something is broken

**Random Baseline for 3 Classes:** Loss  $\approx$  1.099

---

## 1.2 Quick Examples

### 1.2.1 Confident & Correct (Best)

True: Nomenclature

Predicted: [0.95, 0.03, 0.02]  $\rightarrow$  95% confident in Nomenclature

Loss =  $-\log(0.95)$  = 0.05 **✓** Very low!

### 1.2.2 Uncertain & Correct (OK)

True: Nomenclature

Predicted: [0.50, 0.30, 0.20]  $\rightarrow$  50% confident

Loss =  $-\log(0.50)$  = 0.69 **⚠** Higher loss

### 1.2.3 Confident & Wrong (Bad)

True: Nomenclature

Predicted: [0.05, 0.90, 0.05]  $\rightarrow$  90% confident in wrong class!

Loss =  $-\log(0.05)$  = 3.0 **✗** Very high loss

---

## 1.3 Training Health Checklist

### 1.3.1 **✓** Healthy Training

- ☐ Loss decreases each epoch
- ☐ Accuracy increases each epoch
- ☐ Final loss < 0.5
- ☐ Final accuracy > 0.75
- ☐ Training and validation loss both decrease

### 1.3.2 **⚠** Warning Signs

**Overfitting:** - [ ] Training loss keeps decreasing - [ ] Validation loss starts increasing - [ ] Training accuracy » validation accuracy

**Underfitting:** - [ ] Loss stuck near 1.1 (random baseline) - [ ] Minimal improvement per epoch - [ ] Both train and val loss remain high

**Instability:** - [ ] Loss jumps up and down - [ ] Large variations between epochs - [ ] Loss occasionally increases

---

## 1.4 Target Metrics (Your Model)

Metric	Minimum	Good	Excellent
<b>Loss</b>	< 0.7	< 0.5	< 0.3
<b>Accuracy</b>	> 0.70	> 0.80	> 0.90
<b>F1 Score</b>	> 0.65	> 0.75	> 0.85

**Your Current Performance:** 79.9% accuracy  (Good!)

---

## 1.5 When to Stop Training

### 1.5.1 Stop Now



- Final loss < 0.4 and accuracy > 0.80
- Loss hasn't improved for 2+ epochs
- Validation loss increasing
- Accuracy reached target

### 1.5.2 Keep Training

- Loss still decreasing steadily
- Accuracy still improving
- Final loss > 0.5
- No signs of overfitting

---

## 1.6 Loss vs Accuracy Decision Matrix

Loss	Accuracy	Interpretation	Action
0.4	0.80	 Confident, correct	Use model
0.8	0.80	 Uncertain, correct	Train more

Loss	Accuracy	Interpretation	Action
0.2	0.95	⚠️ May be overfitting	Check validation
0.5	0.50	❌ Confident, wrong	Fix model/data
1.0	0.40	❌ Random guessing	Restart training

## 1.7 What Loss Tells You

### Beyond “Lower is Better”:

- Confidence Level:** How sure the model is
  - Loss 0.05 = 95% confident
  - Loss 0.69 = 50% confident
  - Loss 3.0 = 5% confident (but wrong class!)
- Learning Progress:** Fine-grained tracking
  - Accuracy: 0.79 → 0.80 (small change)
  - Loss: 0.65 → 0.48 (significant improvement in confidence)
- Overfitting Detection:** Early warning
  - Training loss: 0.2 ✅
  - Validation loss: 0.8 ⚠️ Overfitting!
- Per-Class Performance:**
  - Nomenclature loss: 0.35 ✅ Easy class
  - Description loss: 0.42 ✅ Easy class
  - Misc loss: 0.85 ⚠️ Hard class

## 1.8 Common Questions

**Q: I have 80% accuracy, is that good?** A: Check the loss! - Loss < 0.5 → Yes, confident and correct ✅ - Loss > 0.8 → Not really, uncertain predictions ⚠️

**Q: Loss went from 0.5 to 0.55, should I worry?** A: Minor fluctuations are normal. Worry if it keeps increasing for 2+ epochs.

**Q: Can I have low loss but low accuracy?** A: Rare, but possible if:  
- Evaluation is broken - Class labels are wrong - Model outputs are not being interpreted correctly


**Q: My loss is 0.3 after 2 epochs, should I keep training?** A: Probably overfitting soon. Check validation loss!

## 1.9 At a Glance

### Excellent Model:

Epoch 4/4 - loss: 0.35 - accuracy: 0.85


Test: loss: 0.38 - accuracy: 0.82

→  Deploy this model!

### Good Model (Your Case):

Epoch 4/4 - loss: 0.48 - accuracy: 0.80

Test: accuracy: 0.7990

→  Ready for production

### Needs Work:

Epoch 4/4 - loss: 0.85 - accuracy: 0.60


Test: accuracy: 0.55

→  Train longer or adjust model

### Overfitting:

Train: loss: 0.15 - accuracy: 0.95

Test: loss: 0.75 - accuracy: 0.70

→  Add dropout, reduce epochs

---

### Keep this card handy during training!

For detailed explanation, see: [understanding\\_rnn\\_training\\_metrics.md](#)