# **Appendix**

We provide supplementary experimental results in this Appendix.

#### 1 HUMAN EVALUATION

To answer the question "which BLEU is correlated with the human perception the most", we conduct the human evaluation. Here we provide the details of the correlation coefficients in this work.

#### 1.1 Human Annotation

Each annotator is asked to assign scores from 0 to 4 to measure the semantic similarity between reference and generated summaries. The meaning of the score is shown in Table 1

#### 1.2 Correlation Coefficient

Kendall's  $\tau$  formulation:

$$\tau = \frac{|Concordant - pairs| - |Discordant - pairs|}{|Concordant - pairs| + |Discordant - pairs|}$$
 (1)

where concordant-pairs are the set where both human scores and automatic metrics suggest the same order for any two generated summaries. Discordant is the set where human assessment disagrees with the order that automatic metrics suggest for any two generated summaries. If two generated summaries are assigned the same BLEU score, then the concordant pair and discordant pair both get a half count.

Pearson correlation coefficient *y* is computed as follow:

$$\gamma = \frac{Cov(X, Y)}{\sqrt{s_X^2 s_y^2}} \tag{2}$$

where X and Y are two continuous random variables. Cov(X,Y) is the covariance.  $s_x^2$  and  $s_y^2$  is the variance.

The Spearman correlation coefficient  $\rho$  between two variables is equal to the Pearson correlation between the rank values of those two variables.

Before the calculation of correlation coefficients, these human direct scores are converted into relative rankings as Direct Assessment Relative Rankings (DaRR) serve as the golden standard for a corpus.

## 1.3 Experiment Result

In the main text, we show the values of correlation coefficient under different size corpus when using arithmetic average to aggregate summary-level human score as corpus-level score. Here, we show the values of correlation coefficient under different size corpus when using geometric average to aggregate summary-level human score as corpus-level score in Table 2. The conclusion is that BLEU-DC, sentence-level BLEU with method $_4$ , is more correlated with human perception.

#### 2 ENSEMBLE MODEL

Leclair et al. [1] proposed two ensemble architecture and prove that the stacking-based technique is better than bagging-based. Thus, We use the stacking-based technique(Figure 1) to aggregate component models that are trained on the whole training set, but

Table 1: The meaning of scores in human evaluation[2].

Score	Meaning
0	No similarity between the generation and reference.
1	Have few shared tokens, not semantically similar.
2	Have some shared tokens, probable semantically similar.
3	Much similar in semantic but a few tokens are different.
4	Identical in semantic.

the training set is processed with different pre-processing. In detail, the ensemble technique just simply averages the last softmax output of each model for every time step when generating the summary.

#### 3 RQ2 EVALUATED WITH OTHER METRICS

• Different pre-processing operations:

Table 3, Table 4, Table 5, Table 6, Table 7, Table 8, Table 9, and Table 10 show the results evaluated with BLEU-DM, BLEU-FC, BLEU-CN, BLEU-NCS, BLEU-RC, Rouge, Meteor, and Cider, respectively.

Statistical tests:

Table 11, Table 12, Table 13, Table 14, Table 15, Table 16, Table 17, Table 18, and Table 19 show the results evaluated with BLEU-DM, BLEU-FC, BLEU-DC, BLEU-CN, BLEU-NCS, BLEU-RC, Rouge, Meteor, and Cider, respectively.

- Different pre-processing operation combinations:
   Table 20, Table 21, Tab 22, Table 23, Table 24, Table 25, Table 26, and Table 27 show the results evaluated with BLEU-DM, BLEU-FC, BLEU-CN, BLEU-NCS, BLEU-RC, Rouge, Meteor, and Cider, respectively.
- In summary:

The findings and conclusions from the above tables are consistent with RQ2 in the main text.

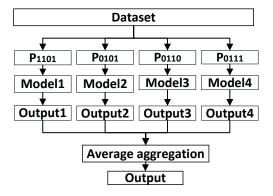


Figure 1: Ensemble Strategy.

Table 2: The values of the correlation coefficients. Here we use geometric average to aggregate summary-level human score as the corpus-level score.

Metric	1		2	20		40		60		80		100	
	τ	$\rho$											
BLEU-DM $s, m_0$	0.32	0.68	0.45	0.63	0.46	0.64	0.46	0.64	0.46	0.64	0.45	0.64	
BLEU-FC $c, m_0$	0.32	0.68	0.31	0.45	0.3	0.44	0.29	0.43	0.29	0.43	0.29	0.42	
BLEU-DC $s, m_4$	0.54	0.75	0.48	0.67	0.49	0.68	0.49	0.68	0.5	0.69	0.49	0.68	
BLEU-CN $s, m_2$	0.47	0.66	0.46	0.64	0.47	0.66	0.48	0.67	0.48	0.67	0.48	0.67	
BLEU-NCS $s, m_l$	0.37	0.53	0.42	0.6	0.44	0.62	0.44	0.62	0.45	0.63	0.44	0.62	
BLEU-RC $s, m_0$	0.32	0.68	0.45	0.63	0.46	0.64	0.46	0.64	0.46	0.64	0.45	0.64	

## 4 RQ3 EVALUATED WITH OTHER METRICS

#### • Corpus sizes:

Table 28, Table 29, Table 31, Table 32, Table 33, Table 34, Table 35, and Table 36 show the results evaluated with BLEU-DM, BLEU-FC, BLEU-CN, BLEU-NCS, BLEU-RC, Rouge, Meteor, and Cider, respectively.

#### • Data splitting ways:

Table 37, Table 38, Table 39, Table 40, Table 41, Table 42, Table 43, and Table 44 show the results evaluated with BLEU-DM, BLEU-FC, BLEU-CN, BLEU-NCS, BLEU-RC, Rouge, Meteor, and Cider, respectively.

### • Duplication ratios:

Table 53, Table 54, Table 30, Table 55, Table 56, Table 57, Table 58, Table 59, and Table 60 show the results evaluated with BLEU-DM, BLEU-FC, BLEU-DC, BLEU-CN, BLEU-NCS, BLEU-RC, Rouge, Meteor, and Cider, respectively.

#### • In summary:

The findings and conclusions from the above tables are consistent with RQ3 in the main text.

# 5 OTHERS

## • Data difference:

Experiments result when we control all three factors (splitting methods, duplication ratios, and dataset sizes): Table 61, Table 62, Table 63, Table 64, Table 65, Table 66, Table 67, Table 68, and Table 69 show the results evaluated with BLEU-DM, BLEU-FC, BLEU-DC, BLEU-CN, BLEU-NCS, BLEU-RC, Rouge, Meteor, and Cider, respectively.

#### • In summary:

We observe that even when we control all three factors (splitting methods, duplication ratios, and dataset sizes), the performance of the same model still varies greatly between different datasets. This indicates that the differences in training data may also be a factor that affects the performance of code summarization. We leave it to future work to study the impact of data differences

Table 3: The result of four code pre-processing operations. Evaluated with BLEU-DM.

Model	$R_0$	$R_1$	$S_0$	$S_1$	$F_0$	$F_1$	$L_0$	$L_1$
CodeNN	4.4	4.38	4.44	4.34	4.39	4.39	4.4	4.38
Astattgru	3.07	3.13	2.95	3.25	3.04	3.16	2.99	3.2
Rencos	19.21	18.9	18.35	19.75	19.17	18.94	18.77	19.34
NCS	8.9	8.74	8.5	9.14	8.74	8.9	8.51	9.13
Avg.	8.90	8.78	8.56	9.12	8.84	8.85	8.67	9.01

Table 4: The result of four code pre-processing operations. Evaluated with BLEU-FC.

Model	$R_0$	$R_1$	$S_0$	$S_1$	$F_0$	$F_1$	$L_0$	$L_1$
CodeNN	5.13	5.16	5.12	5.17	5.13	5.16	5.15	5.14
Astattgru	4.48	4.52	4.34	4.66	4.43	4.57	4.38	4.62
Rencos	20.79	20.56	20.01	21.34	20.85	20.5	20.33	21.02
NCS	7.69	7.48	7.32	7.85	7.47	7.7	7.66	7.51
Avg.	9.52	9.43	9.2	9.76	9.47	9.48	9.38	9.57

Table 5: The result of four code pre-processing operations. Evaluated with BLEU-CN.

Model	$R_0$	$R_1$	$S_0$	$S_1$	$F_0$	$F_1$	$L_0$	$L_1$
CodeNN	13.50	13.51	13.39	13.62	13.51	13.51	13.52	13.49
Astattgru	12.66	12.64	12.13	13.17	12.55	12.75	12.50	12.80
Rencos	27.6	27.37	26.57	28.41	27.53	27.44	27.24	27.73
NCS	19.52	19.43	18.8	20.15	19.37	19.57	19.16	19.79
Avg.	18.32	18.24	17.72	18.84	18.24	18.32	18.10	18.45

Table 6: The result of four code pre-processing operations. Evaluated with BLEU-NCS.

Model	$R_0$	$R_1$	$S_0$	$S_1$	$F_0$	$F_1$	$L_0$	$L_1$
CodeNN	14.45	14.46	14.44	14.48	14.46	14.46	14.47	14.45
Astattgru	13.48	13.49	13.07	13.9	13.41	13.56	13.35	13.62
Rencos	28.24	28.0	27.3	28.94	28.19	28.05	27.88	28.36
NCS	19.95	19.88	19.43	20.4	19.83	20.0	19.63	20.19
Avg.	19.03	18.96	18.56	19.43	18.97	19.02	18.83	19.16

Table 7: The result of four code pre-processing operations. Evaluated with BLEU-RC.

Model	$R_0$	$R_1$	$S_0$	$S_1$	$F_0$	$F_1$	$L_0$	$L_1$
CodeNN	4.4	4.38	4.44	4.34	4.39	4.4	4.4	4.38
Astattgru	3.07	3.13	2.95	3.25	3.04	3.16	2.99	3.2
Rencos	19.21	18.9	18.35	19.75	19.17	18.94	18.77	19.34
NCS	8.9	8.74	8.5	9.14	8.74	8.9	8.51	9.13
Avg.	8.9	8.79	8.56	9.12	8.84	8.85	8.67	9.01

Table 8: The result of four code processing operations. Evaluated with Rouge.

Model	$R_0$	$R_1$	$S_0$	$S_1$	$F_0$	$F_1$	$L_0$	$L_1$
CodeNN	8.42	8.47	8.1	8.79	8.39	8.5	8.44	8.45
Astattgru	7.4	7.37	6.85	7.93	7.3	7.47	7.21	7.56
Rencos	16.63	16.53	15.78	17.37	16.64	16.51	16.38	16.77
NCS	11.16	11.07	10.37	11.86	11.02	11.21	11.0	11.24
Avg.	10.9	10.86	10.27	11.49	10.84	10.92	10.76	11.0

Table 9: The result of four code pre-processing operations. Evaluated with Meteor.

Model	$R_0$	$R_1$	$S_0$	$S_1$	$F_0$	$F_1$	$L_0$	$L_1$
CodeNN	24.8	24.95	24.32	25.44	24.84	24.91	24.87	24.88
Astattgru	24.74	24.63	23.33	26.04	24.45	24.92	24.44	24.93
Rencos	39.68	39.53	38.04	41.17	39.49	39.72	39.36	39.86
NCS	33.68	33.72	32.02	35.37	33.56	33.83	33.34	34.06
Avg.	30.72	30.71	29.43	32.01	30.58	30.84	30.5	30.93

Table 10: The result of four code pre-processing operations. Evaluated with Cider.

Model	$R_0$	$R_1$	$S_0$	$S_1$	$F_0$	$F_1$	$L_0$	$L_1$
CodeNN	0.68	0.68	0.67	0.69	0.68	0.68	0.68	0.68
Astattgru	0.5	0.5	0.45	0.55	0.49	0.51	0.48	0.52
Rencos	2.07	2.04	1.96	2.15	2.07	2.04	2.03	2.08
NCS	1.21	1.21	1.12	1.3	1.2	1.22	1.16	1.26
Avg.	1.12	1.11	1.05	1.17	1.11	1.11	1.09	1.14

Table 11: p-values of the t-test and Wilcoxon-Mann-Whitney-test (WMW-test). Evaluated with BLEU-DM.

Model		t-t	est		WMW-test				
	R	S	F	L	R	S	F	L	
CodeNN	0.7737	0.2626	0.9229	0.7737	0.9581	0.3184	0.7929	0.7132	
Astattgru	0.6821	0.0344	0.4273	0.1477	0.7929	0.0661	0.4309	0.1893	
Rencos	0.5690	0.0043	0.6780	0.2855	0.7929	0.0054	0.1893	0.4309	
NCS	0.5942	0.0188	0.5859	0.0239	0.9581	0.0239	0.9581	0.0406	

Table 12: p-values of the t-test and Wilcoxon-Mann-Whitney-test (WMW-test). Evaluated with BLEU-FC.

Model		t-t	est		WMW-test				
	R	S	F	L	R	S	F	L	
CodeNN	0.7675	0.5573	0.7512	0.9271	0.7132	0.9581	0.8748	0.9581	
Astattgru	0.8811	0.1200	0.5286	0.2466	0.7929	0.1563	0.7132	0.2701	
Rencos	0.6869	0.0089	0.5410	0.2121	0.7132	0.0136	0.1036	0.2271	
NCS	0.4474	0.0309	0.3737	0.5585	0.4309	0.0520	0.4309	0.4309	

Table 13: p-values of the t-test and Wilcoxon-Mann-Whitney-test (WMW-test). Evaluated with BLEU-DC.

Model		t-t	est			WMV	V-test	
Model	R	S	F	L	R	S	F	L
CodeNN	0.9272	0.8950	0.9098	0.8370	0.9581	0.7929	0.9581	0.9581
Astattgru	0.7846	0.0003	0.4086	0.2233	0.9581	0.0028	0.4948	0.3184
Rencos	0.6010	0.0014	0.7698	0.3267	0.7929	0.0009	0.4309	0.5635
NCS	0.7581	0.0019	0.5814	0.0731	0.9581	0.0014	0.9581	0.1893

Table 14: p-values of the t-test and Wilcoxon-Mann-Whitney-test (WMW-test). Evaluated with BLEU-CN.

Model		t-t	est			WMV	V-test	
Model	R	S	F	L	R	S	F	L
CodeNN	0.9366	0.0750	0.9999	0.8018	0.7132	0.0406	0.8748	0.9581
Astattgru	0.9669	0.0000	0.5421	0.3448	0.7929	0.0009	0.6365	0.4948
Rencos	0.6953	0.0003	0.8723	0.4093	0.7929	0.0009	0.4948	0.6365
NCS	0.8396	0.0002	0.6579	0.1452	0.9581	0.0009	0.7929	0.2271

Table 15: p-values of the t-test and Wilcoxon-Mann-Whitney-test (WMW-test). Evaluated with BLEU-NCS.

Model		t-t	est			WMV	V-test	
Model	R	S	F	L	R	S	F	L
CodeNN	0.9078	0.6509	0.9581	0.8344	0.8748	0.9581	0.8748	0.9581
Astattgru	0.9720	0.0001	0.5544	0.2888	0.8748	0.0009	0.6365	0.4948
Rencos	0.6681	0.0006	0.8009	0.3774	0.8748	0.0009	0.4309	0.6365
NCS	0.8283	0.0010	0.6285	0.0939	0.9581	0.0009	0.8748	0.2271

Table 16: p-values of the t-test and Wilcoxon-Mann-Whitney-test (WMW-test). Evaluated with BLEU-RC.

Model		t-t	est			WMV	V-test	
1,10001	R	S	F	L	R	S	F	L
CodeNN	0.7730	0.2609	0.9214	0.7746	0.9581	0.3184	0.7929	0.7132
Astattgru	0.6827	0.0345	0.4270	0.1476	0.7929	0.0661	0.4309	0.1893
Rencos	0.5690	0.0043	0.6780	0.2856	0.7929	0.0054	0.1893	0.4309
NCS	0.5944	0.0188	0.5858	0.0239	0.9581	0.0239	0.9581	0.0406

Table 17: p-values of the t-test and Wilcoxon-Mann-Whitney-test (WMW-test). Evaluated with Rouge.

Model		t-t	est			WMV	V-test	
1,10 de1	R	S	F	L	R	S	F	L
CodeNN	0.7348	0.0023	0.8873	0.9746	0.6365	0.0136	0.8748	0.9581
Astattgru	0.8902	0.0000	0.5605	0.5421	0.7929	0.0009	0.5635	0.5635
Rencos	0.8698	0.0000	0.7912	0.5752	0.5635	0.0009	0.9581	0.4948
NCS	0.9716	0.0000	0.7831	0.4644	0.9581	0.0009	0.6365	0.4948

Table 18: p-values of the t-test and Wilcoxon-Mann-Whitney-test (WMW-test). Evaluated with Meteor.

Model		t-t	est			WMV	V-test	
1710 461	R	S	F	L	R	S	F	L
CodeNN	0.8542	0.0011	0.6720	0.9564	0.7130	0.0117	0.2476	0.9163
Astattgru	0.9256	0.0000	0.6124	0.2802	0.7929	0.0009	0.4948	0.2701
Rencos	0.8363	0.0000	0.7848	0.4200	0.8748	0.0009	0.4309	0.4948
NCS	0.8250	0.0000	0.6639	0.5793	0.8747	0.0009	0.5283	0.5632

Table 19: p-values of the t-test and Wilcoxon-Mann-Whitney-test (WMW-test). Evaluated with Cider.

Model		t-t	est			WMV	V-test	
1710 461	R	S	F	L	R	S	F	L
CodeNN	1.0000	0.0582	0.8233	0.8233	0.8720	0.0763	1.0000	0.9145
Astattgru	0.9113	0.0002	0.6282	0.2405	1.0000	0.0009	0.5984	0.2466
Rencos	0.6509	0.0002	0.7104	0.3733	0.7525	0.0009	0.4306	0.5283
NCS	0.9018	0.0003	0.6500	0.0785	0.9580	0.0009	0.6733	0.1138

Table 20: Performance of different code pre-processing combinations. Evaluated with BLEU-DM.

Model	$P_{0000}$	$P_{0001}$	$P_{0010}$	$P_{0011}$	$P_{0100}$	$P_{0101}$	$P_{0110}$	$P_{0111}$	$P_{1000}$	$P_{1001}$	$P_{1010}$	$P_{1011}$	$P_{1100}$	$P_{1101}$	$P_{1110}$	$P_{1111}$	Ensemble
CodeNN	4.29	4.41	4.24	4.55	4.7	4.08	4.56	4.17	4.55	4.45	4.5	4.55	4.18	4.49	4.21	4.3	7.89
Astattgru	3.01	2.83	2.75	2.85	3.3	3.52	3.24	3.53	3.15	3.07	2.66	3.27	2.81	3.57	3.01	2.99	8.60
Rencos																	
NCS	7.99	8.85	7.94	8.96	8.53	9.69	8.69	9.25	8.62	8.41	8.66	8.55	8.95	10.14	8.67	9.16	16.84

Table 21: Performance of different code pre-processing combinations. Evaluated with BLEU-FC.

Model	$P_{0000}$	$P_{0001}$	$P_{0010}$	$P_{0011}$	$P_{0100}$	$P_{0101}$	$P_{0110}$	$P_{0111}$	$P_{1000}$	$P_{1001}$	$P_{1010}$	$P_{1011}$	$P_{1100}$	$P_{1101}$	$P_{1110}$	$P_{1111}$	Ensemble
CodeNN	5.12	5.18	4.9	5.14	5.48	5.03	5.34	5.06	5.15	4.98	5.11	5.38	5.0	5.31	5.08	5.04	8.28
Astattgru	4.56	4.01	3.84	4.13	4.95	4.8	4.73	5.1	4.76	4.63	3.78	4.97	4.11	4.73	4.27	4.62	10.32
Rencos																	
NCS	7.02	7.26	7.32	7.11	7.68	8.11	7.52	7.84	8.17	6.9	7.95	6.83	7.85	8.63	7.81	7.37	16.35

Table 22: Performance of different code pre-processing combinations, evaluated with BLEU-CN.

Model	$P_{0000}$	$P_{0001}$	$P_{0010}$	$P_{0011}$	$P_{0100}$	$P_{0101}$	$P_{0110}$	$P_{0111}$	$P_{1000}$	$P_{1001}$	$P_{1010}$	$P_{1011}$	$P_{1100}$	$P_{1101}$	$P_{1110}$	$P_{1111}$	Ensemble
CodeNN	13.21	13.29	13.21	13.36	14.10	13.44	13.84	13.64	13.36	13.41	13.46	13.83	13.53	13.71	13.47	13.23	16.82
Astattgru	11.98	12.40	12.08	11.98	12.74	13.77	13.10	13.09	12.25	12.43	11.99	11.92	12.74	13.68	13.12	13.13	17.36
Rencos	26.04	26.08	26.93	26.61	27.54	29.20	28.16	28.40	26.59	26.64	26.78	26.87	27.44	29.98	28.46	28.06	30.01
NCS	18.47	19.09	18.23	19.11	19.62	20.83	19.65	20.42	19.06	18.68	18.97	18.76	19.70	21.11	19.56	20.30	26.15

Table 23: Performance of different code pre-processing combinations, Evaluated with BLEU-NCS.

Model	$P_{0000}$	$P_{0001}$	$P_{0010}$	$P_{0011}$	$P_{0100}$	$P_{0101}$	$P_{0110}$	$P_{0111}$	$P_{1000}$	$P_{1001}$	$P_{1010}$	$P_{1011}$	$P_{1100}$	$P_{1101}$	$P_{1110}$	$P_{1111}$	Ensemble
CodeNN	14.27	14.43	14.37	14.38	14.84	14.25	14.67	14.5	14.43	14.51	14.47	14.64	14.37	14.58	14.31	14.3	17.65
Astattgru																	
Rencos	26.73	26.8	27.71	27.4	28.07	29.67	28.68	28.93	27.31	27.31	27.51	27.64	27.97	30.52	29.03	28.61	30.54
NCS	19.1	19.71	18.95	19.78	19.9	21.01	19.94	20.61	19.62	19.28	19.59	19.41	20.05	21.29	19.91	20.44	26.52

Table 24: Performance of different code pre-processing combinations. Evaluated with BLEU-RC.

Model	$P_{0000}$	$P_{0001}$	$P_{0010}$	$P_{0011}$	$P_{0100}$	$P_{0101}$	$P_{0110}$	$P_{0111}$	$P_{1000}$	$P_{1001}$	$P_{1010}$	$P_{1011}$	$P_{1100}$	$P_{1101}$	$P_{1110}$	$P_{1111}$	Ensemble
CodeNN	4.29	4.41	4.24	4.55	4.71	4.08	4.56	4.17	4.55	4.45	4.5	4.55	4.18	4.49	4.21	4.3	7.89
Astattgru	3.01	2.83	2.75	2.85	3.3	3.52	3.24	3.53	3.15	3.07	2.66	3.27	2.81	3.57	3.01	2.99	8.60
Rencos																	
NCS	7.99	8.85	7.94	8.96	8.53	9.69	8.69	9.25	8.62	8.41	8.66	8.55	8.95	10.14	8.67	9.16	16.84

Table 25: Performance of different code pre-processing combinations. Evaluated with Rouge.

Model	$P_{0000}$	$P_{0001}$	$P_{0010}$	$P_{0011}$	$P_{0100}$	$P_{0101}$	$P_{0110}$	$P_{0111}$	$P_{1000}$	$P_{1001}$	$P_{1010}$	$P_{1011}$	$P_{1100}$	$P_{1101}$	$P_{1110}$	$P_{1111}$	Ensemble
CodeNN	24.21	24.11	24.02	24.14	26.19	25.53	25.81	25.58	24.0	24.05	24.2	25.8	25.49	25.67	25.03	24.18	28.92
Astattgru	22.92	23.98	23.13	23.07	25.11	27.26	26.15	25.41	23.71	23.96	23.19	22.66	25.47	26.93	25.84	26.16	29.42
Rencos	37.85	37.9	38.18	37.87	40.48	41.99	40.76	41.23	38.13	38.37	38.02	38.04	40.5	42.57	40.93	40.87	42.61
NCS	31.86	32.14	31.21	32.0	34.9	36.55	34.83	36.24	32.72	31.96	32.52	31.78	34.2	36.34	34.49	35.44	40.34

Table 26: Performance of different code pre-processing combinations. Evaluated with Meteor.

Model	$P_{0000}$	$P_{0001}$	$P_{0010}$	$P_{0011}$	$P_{0100}$	$P_{0101}$	$P_{0110}$	$P_{0111}$	$P_{1000}$	$P_{1001}$	$P_{1010}$	$P_{1011}$	$P_{1100}$	$P_{1101}$	$P_{1110}$	$P_{1111}$	Ensemble
CodeNN	8.14	8.09	7.82	7.99	9.06	8.96	8.84	8.85	8.01	7.92	7.95	8.91	8.92	8.89	8.77	8.01	10.11
Astattgru	6.82	6.91	6.5	6.77	7.71	8.43	8.15	7.68	6.99	7.14	6.52	7.13	7.48	8.29	7.51	8.16	11.10
Rencos	15.39	15.51	15.97	15.87	16.72	17.92	17.29	17.54	15.69	15.81	15.95	16.06	16.71	18.33	17.34	17.12	18.35
NCS	10.2	10.45	10.19	10.38	11.63	12.31	11.37	12.01	10.73	10.28	10.64	10.1	11.6	12.48	11.6	11.88	15.79

Table 27: Performance of different code pre-processing combinations. Evaluated with Cider.

Model	$P_{0000}$	$P_{0001}$	$P_{0010}$	$P_{0011}$	$P_{0100}$	$P_{0101}$	$P_{0110}$	$P_{0111}$	$P_{1000}$	$P_{1001}$	$P_{1010}$	$P_{1011}$	$P_{1100}$	$P_{1101}$	$P_{1110}$	P <sub>1111</sub>	Ensemble
CodeNN	0.65	0.66	0.65	0.67	0.72	0.67	0.7	0.69	0.67	0.66	0.67	0.7	0.69	0.7	0.67	0.65	0.99
Astattgru	0.44	0.46	0.44	0.43	0.52	0.62	0.56	0.54	0.45	0.47	0.42	0.46	0.49	0.61	0.52	0.56	1.07
Rencos																	
NCS	1.08	1.16	1.05	1.17	1.22	1.4	1.22	1.35	1.12	1.13	1.11	1.12	1.25	1.43	1.22	1.33	1.97

Table 28: Performance in Different datasets. Evaluated with BLEU-DM.

		Dataset	
Model	TLC	FCM	CSN
CodeNN	26.23±0.15	7.26±0.14	1.56±0.11
Deepcom	$13.38 \pm 2.00$	$4.95 \pm 0.04$	$0.99 \pm 0.25$
Astattgru	$24.09 \pm 0.74$	$9.92 \pm 0.06$	$5.14 \pm 4.05$
Rencos	$40.40 \pm 0.01$	$10.40 \pm 0.00$	$4.76 \pm 0.06$
NCS	$37.30 \pm 0.23$	$12.26 \pm 0.54$	$3.13 \pm 0.53$

Table 29: Performance in Different datasets. Evaluated with BLEU-FC.

Model		Dataset	
	TLC	FCM	CSN
CodeNN	25.63±0.47	11.99±0.21	$2.75 \pm 0.02$
Deepcom	11.19±2.89	$8.73 \pm 0.09$	$1.52 \pm 0.45$
Astattgru	$25.21 \pm 0.96$	$15.94 \pm 0.09$	$4.87 \pm 0.91$
Rencos	$41.20 \pm 0.17$	$16.34 \pm 0.00$	$7.29 \pm 0.16$
NCS	$34.74 \pm 0.83$	$17.72 \pm 0.74$	$2.11 \pm 1.65$

Table 30: Performance in Different datasets. Evaluated with BLEU-DC

Model		Dataset	
1110 0001	TLC	FCM	CSN
CodeNN	28.24±0.19	12.64±0.13	3.32±0.09
Deepcom	$15.65 \pm 2.12$	$9.12 \pm 0.03$	$1.98 \pm 0.30$
Astattgru	$25.90 \pm 0.79$	$15.58 \pm 0.11$	$6.86 \pm 3.07$
Rencos	$42.46 \pm 0.05$	$15.47 \pm 0.00$	$6.65 \pm 0.05$
NCS	$39.50 \pm 0.23$	$18.07 \pm 0.46$	$6.66 \pm 0.51$

Table 31: Performance on different datasets. Evaluated with BLEU-CN.

	Dataset	
TLC	FCM	CSN
33.03±0.20	25.26±0.01	8.58±0.15
$20.54 \pm 2.57$	$20.80 \pm 0.02$	$6.12 \pm 0.64$
$30.19 \pm 0.86$	$27.63 \pm 0.24$	$11.73 \pm 0.41$
$46.81 \pm 0.06$	$25.82 \pm 0.00$	$11.19 \pm 0.09$
$44.25 \pm 0.21$	$30.69 \pm 0.12$	$11.80 \pm 0.94$
	33.03±0.20 20.54±2.57 30.19±0.86 46.81±0.06	TLC FCM  33.03±0.20 25.26±0.01 20.54±2.57 20.80±0.02 30.19±0.86 27.63±0.24 46.81±0.06 25.82±0.00

 ${\bf Table~32: Performance~in~Different~datasets.~Evaluated~with~BLEU-NCS.}$ 

Model		Dataset	
1110461	TLC	FCM	CSN
CodeNN	33.73±0.19	26.43±0.06	11.40±0.32
Deepcom	$21.63 \pm 2.43$	$23.44 \pm 0.06$	$9.18\pm0.30$
Astattgru	$30.92 \pm 0.81$	$28.44 \pm 0.24$	12.44±1.14
Rencos	$47.23 \pm 0.05$	$28.23 \pm 0.00$	$14.76 \pm 0.08$
NCS	$44.55 \pm 0.22$	$31.18 \pm 0.34$	$14.68 \pm 0.62$

Table 33: Performance in Different datasets. Evaluated with BLEU-RC.

Model	Dataset					
1,10001	TLC	FCM	CSN			
CodeNN	26.23±0.15	7.28±0.14	1.56±0.11			
Deepcom	$13.38 \pm 2.00$	$4.96 \pm 0.04$	$0.99 \pm 0.25$			
Astattgru	$24.09 \pm 0.74$	$9.94 \pm 0.06$	$5.14 \pm 4.05$			
Rencos	$40.40 \pm 0.01$	$10.41 \pm 0.00$	$4.76 \pm 0.06$			
NCS	$37.30 \pm 0.23$	$12.30 \pm 0.53$	$3.13 \pm 0.53$			

Table 34: Performance in Different datasets. Evaluated with Rouge.

Model	Dataset					
1110401	TLC	FCM	CSN			
CodeNN	42.67±0.34	34.98±0.05	16.01±0.38			
Deepcom	$30.44 \pm 3.48$	$28.13 \pm 0.05$	$10.86 \pm 0.23$			
Astattgru	$39.07 \pm 0.99$	$38.59 \pm 0.22$	$22.20 \pm 1.76$			
Rencos	$56.45 \pm 0.08$	$35.78 \pm 0.00$	$20.78 \pm 0.10$			
NCS	$54.68 \pm 0.22$	40.61±0.19	21.91±2.34			

Table 35: Performance in Different datasets. Evaluated with Meteor.

Model		Dataset	
	TLC	FCM	CSN
CodeNN	19.24±0.23	14.84±0.21	5.36±0.26
Deepcom	$12.10 \pm 1.33$	$10.42 \pm 0.01$	$3.72 \pm 0.35$
Astattgru	$17.52 \pm 0.60$	$17.72 \pm 0.10$	$8.68 \pm 1.01$
Rencos	$28.63 \pm 0.05$	$16.55 \pm 0.00$	$8.94 \pm 0.08$
NCS	$26.22 \pm 0.30$	19.04±0.25	$6.39 \pm 1.19$

Table 36: Performance in Different datasets. Evaluated with Cider.

Model	Dataset					
1110 4101	TLC	FCM	CSN			
CodeNN	2.74±0.02	1.36±0.01	0.32±0.02			
Deepcom	$1.46 \pm 0.22$	$0.76 \pm 0.01$	$0.15 \pm 0.03$			
Astattgru	$2.44 \pm 0.08$	$1.78 \pm 0.02$	$0.36 \pm 0.19$			
Rencos	$4.12 \pm 0.01$	$1.63 \pm 0.00$	$0.58 \pm 0.01$			
NCS	$3.89 \pm 0.02$	$2.21 \pm 0.03$	$0.63 \pm 0.06$			

Table 37: The result in different corpus sizes. Evaluated with BLEU-DM.

Model	$FCM_{Method-Small}$	$FCM_{Method\text{-}Medium}$	$FCM_{Method-Large}$	${\rm CSN}_{\rm Method\text{-}Small}$	CSN <sub>Method-Medium</sub>
CodeNN	5.31±0.17	$9.10\pm0.15$	$12.60 \pm 0.25$	$3.28 \pm 0.06$	$9.26 \pm 0.25$
Deepcom	$4.86 \pm 0.05$	$6.35 \pm 0.16$	$6.96 \pm 0.36$	$5.76 \pm 0.75$	$5.78 \pm 0.78$
Astattgru	$7.80 \pm 0.51$	$12.41 \pm 0.06$	$15.76 \pm 0.08$	$3.93 \pm 0.08$	$12.31 \pm 0.22$
Rencos	$9.48 \pm 0.17$	$16.78 \pm 0.08$	$18.07 \pm 0.05$	$5.61 \pm 0.07$	$16.71 \pm 0.04$
NCS	$8.98 \pm 0.09$	$17.04 \pm 0.31$	$22.93 \pm 0.27$	$5.16 \pm 0.13$	$18.66 \pm 0.28$

Table 38: The result in different corpus sizes. Evaluated with BLEU-FC.

Model	$FCM_{Method\text{-}Small}$	$FCM_{Method-Medium}$	$FCM_{Method-Large}$	$CSN_{Method-Small}$	CSN <sub>Method-Medium</sub>
CodeNN	$9.70 \pm 0.23$	$14.54 \pm 0.27$	19.01±0.30	$3.95 \pm 0.05$	9.88±0.18
Deepcom	$8.23 \pm 0.09$	$10.41 \pm 0.28$	$11.34 \pm 0.54$	$3.34 \pm 0.03$	$3.77 \pm 0.75$
Astattgru	$12.93 \pm 0.71$	$18.84 \pm 0.06$	$22.91 \pm 0.08$	$4.79 \pm 0.45$	$11.85 \pm 0.28$
Rencos	$14.86 \pm 0.15$	$23.65 \pm 0.10$	$25.57 \pm 0.05$	$7.35 \pm 0.52$	$16.26 \pm 0.16$
NCS	$13.66 \pm 0.40$	$23.35 \pm 0.56$	$29.64 \pm 0.54$	$2.61 \pm 0.14$	$11.39 \pm 0.75$

Table 39: The result in different corpus sizes. Evaluated with BLEU-CN.

Model	$FCM_{Method-Small}$	$FCM_{Method\text{-}Medium}$	$FCM_{Method\text{-}Large}$	${\rm CSN}_{\rm Method\text{-}Small}$	CSN <sub>Method-Medium</sub>
CodeNN	22.85±0.12	27.38±0.04	31.19±0.17	9.38±0.14	20.13±0.34
Deepcom	$20.49 \pm 0.16$	$22.78 \pm 0.12$	$23.72 \pm 0.32$	$12.30 \pm 0.64$	$12.64 \pm 1.07$
Astattgru	$23.94 \pm 0.67$	$29.83 \pm 0.20$	$33.36 \pm 0.16$	$11.38 \pm 0.42$	$24.11 \pm 0.25$
Rencos	$24.02 \pm 0.03$	$31.47 \pm 0.04$	$33.95 \pm 0.03$	$11.73 \pm 0.16$	$25.03 \pm 0.02$
NCS	$27.89 \pm 0.37$	$35.41 \pm 0.20$	$40.73\pm0.16$	$12.74 \pm 0.13$	$30.12 \pm 0.27$
OOV Ratio of Deepcom	91.90%	88.94%	88.32%	91.49%	85.81%
OOV Ratio of Others	63.36%	53.09%	48.60%	60.99%	34.00%

Table 40: The result in different corpus sizes. Evaluated with BLEU-NCS.

Model	FCM <sub>Method-Small</sub>	$FCM_{Method-Medium}$	FCM <sub>Method-Large</sub>	CSN <sub>Method-Small</sub>	CSN <sub>Method-Medium</sub>
CodeNN	$24.45 \pm 0.09$	$28.40 \pm 0.10$	$31.96 \pm 0.19$	$12.57 \pm 0.13$	$21.63 \pm 0.17$
Deepcom	$23.18 \pm 0.06$	$24.92 \pm 0.20$	$25.62 \pm 0.37$	$14.49 \pm 0.23$	$14.96 \pm 0.95$
Astattgru	$25.53 \pm 0.64$	$30.60 \pm 0.14$	$33.92 \pm 0.14$	$12.99 \pm 0.13$	$24.05 \pm 0.34$
Rencos	$26.83 \pm 0.08$	$33.98 \pm 0.05$	$35.91 \pm 0.02$	$15.04 \pm 0.16$	$27.64 \pm 0.00$
NCS	$28.29 \pm 0.27$	$35.55 \pm 0.29$	$40.71 \pm 0.28$	$16.57 \pm 0.16$	$32.55 \pm 0.27$

Table 41: The result in different corpus sizes. Evaluated with BLEU-RC.

Model	FCM <sub>Method-Small</sub>	FCM <sub>Method-Medium</sub>	$FCM_{Method-Large}$	CSN <sub>Method-Small</sub>	CSN <sub>Method-Medium</sub>
CodeNN	$5.32 \pm 0.17$	9.12±0.15	$12.64 \pm 0.25$	$3.28 \pm 0.06$	9.28±0.25
Deepcom	$4.87 \pm 0.05$	$6.36 \pm 0.16$	$6.97 \pm 0.36$	$5.76 \pm 0.75$	$5.79 \pm 0.78$
Astattgru	$7.82 \pm 0.51$	$12.44 \pm 0.06$	$15.80 \pm 0.09$	$3.93 \pm 0.08$	$12.32 \pm 0.22$
Rencos	$9.49 \pm 0.17$	$16.82 \pm 0.08$	$18.11 \pm 0.04$	$5.62 \pm 0.07$	$16.73 \pm 0.04$
NCS	$9.02 \pm 0.09$	$17.10 \pm 0.31$	$23.00 \pm 0.27$	$5.16 \pm 0.13$	$18.69 \pm 0.28$

Table 42: The result in different corpus sizes. Evaluated with Rouge.

Model	$FCM_{Method-Small}$	$FCM_{Method\text{-}Medium}$	$FCM_{Method-Large}$	${\rm CSN}_{\rm Method\text{-}Small}$	CSN <sub>Method-Medium</sub>
CodeNN	$32.17 \pm 0.22$	$37.83 \pm 0.16$	$42.43 \pm 0.29$	$16.53 \pm 0.15$	$30.24 \pm 0.29$
Deepcom	$27.68 \pm 0.15$	$30.79 \pm 0.36$	32.04±0.61	$18.64 \pm 0.35$	$19.00 \pm 0.87$
Astattgru	$34.08 \pm 1.02$	$41.13 \pm 0.23$	44.94±0.15	$18.37 \pm 0.11$	$34.21 \pm 0.64$
Rencos	$33.56 \pm 0.03$	$41.86 \pm 0.05$	$45.37 \pm 0.04$	$19.87 \pm 0.20$	$37.04 \pm 0.11$
NCS	$37.83 \pm 0.86$	$45.67 \pm 0.31$	$50.89 \pm 0.33$	$21.22 \pm 0.29$	$40.97 \pm 0.52$

Table 43: The result in different corpus sizes. Evaluated with Meteor.

Model	$FCM_{Method\text{-}Small}$	$FCM_{Method-Medium}$	$FCM_{Method-Large}$	$CSN_{Method-Small}$	CSN <sub>Method-Medium</sub>
CodeNN	$12.93 \pm 0.20$	$16.30 \pm 0.15$	$19.01 \pm 0.21$	$5.95 \pm 0.04$	$11.53 \pm 0.12$
Deepcom	$9.97 \pm 0.01$	$11.67 \pm 0.16$	$12.48 \pm 0.28$	$7.47 \pm 0.25$	$7.47 \pm 0.25$
Astattgru	$14.87 \pm 0.64$	$19.04 \pm 0.09$	$21.40 \pm 0.07$	$6.37 \pm 0.20$	$13.20 \pm 0.04$
Rencos	$15.00\pm0.03$	$20.00 \pm 0.05$	$21.84 \pm 0.03$	$7.67 \pm 0.21$	$15.48 \pm 0.11$
NCS	$16.71 \pm 0.33$	$21.91 \pm 0.22$	$25.25 \pm 0.23$	$6.97 \pm 0.07$	$15.62 \pm 0.24$

Table 44: The result in different corpus sizes. Evaluated with Cider.

Model	$FCM_{Method\text{-}Small}$	$FCM_{Method\text{-}Medium}$	$FCM_{Method-Large}$	${\rm CSN}_{\rm Method\text{-}Small}$	CSN <sub>Method-Medium</sub>
CodeNN	$0.99 \pm 0.02$	$1.50 \pm 0.02$	1.95±0.03	$0.44 {\pm} 0.01$	$1.21 \pm 0.03$
Deepcom	$0.69 \pm 0.02$	$0.91 \pm 0.02$	$1.01 \pm 0.04$	$0.67 \pm 0.08$	$0.68 \pm 0.09$
Astattgru	$1.25 \pm 0.10$	$2.00 \pm 0.02$	$2.41 \pm 0.02$	$0.45 \pm 0.00$	$1.46 \pm 0.03$
Rencos	$1.37 \pm 0.01$	$2.21 \pm 0.01$	$2.48 \pm 0.00$	$0.64 \pm 0.01$	$1.82 \pm 0.01$
NCS	$1.82 \pm 0.01$	$2.70 \pm 0.02$	$3.27 \pm 0.02$	$0.75 \pm 0.02$	$2.20 \pm 0.03$

Table 45: The result in different data splitting methods. Evaluated with BLEU-DM.

Model	CSN <sub>Project-Medium</sub>	CSN <sub>Class-Medium</sub>	CSN <sub>Method-Medium</sub>	$FCM_{Project-Large}$	$FCM_{Method-Large}$
CodeNN	1.56±0.11	$6.82 \pm 0.08$	9.26±0.25	$7.26 \pm 0.14$	$12.60 \pm 0.25$
Deepcom	$0.99 \pm 0.25$	$4.31 \pm 0.22$	$5.78 \pm 0.78$	$4.95 \pm 0.04$	$6.96 \pm 0.36$
Astattgru	$5.14 \pm 4.05$	$8.55 \pm 0.41$	$12.31 \pm 0.22$	$9.92 \pm 0.06$	$15.76 \pm 0.08$
Rencos	$4.76 \pm 0.06$	$11.85 \pm 0.04$	$16.71 \pm 0.04$	$10.40 \pm 0.00$	$18.07 \pm 0.05$
NCS	$3.13 \pm 0.53$	$12.25 \pm 0.14$	$18.66 \pm 0.28$	$12.26 \pm 0.54$	$22.93 \pm 0.27$

Table 46: The result in different data splitting methods. Evaluated with BLEU-FC.

Model	CSN <sub>Project-Medium</sub>	CSN <sub>Class-Medium</sub>	CSN <sub>Method-Medium</sub>	FCM <sub>Project-Large</sub>	FCM <sub>Method-Large</sub>
CodeNN	$2.75 \pm 0.02$	$7.59 \pm 0.17$	9.88±0.18	$11.99 \pm 0.21$	19.01±0.30
Deepcom	$1.52 \pm 0.45$	$3.37 \pm 0.14$	$3.77 \pm 0.75$	$8.73 \pm 0.09$	$11.34 \pm 0.54$
Astattgru	$4.87 \pm 0.91$	$8.60 \pm 0.65$	$11.85 \pm 0.28$	15.94±0.09	$22.91 \pm 0.08$
Rencos	$7.29 \pm 0.16$	$12.20 \pm 0.14$	$16.26 \pm 0.16$	$16.34 \pm 0.00$	$25.57 \pm 0.05$
NCS	$2.11 \pm 1.65$	$7.46 \pm 0.30$	$11.39 \pm 0.75$	$17.72 \pm 0.74$	$29.64 \pm 0.54$

Table 47: The result of different data splitting ways. Evaluated with BLEU-CN.

Model	CSN <sub>Project-Medium</sub>	CSN <sub>Class-Medium</sub>	CSN <sub>Method-Medium</sub>	FCM <sub>Project-Large</sub>	$FCM_{Method-Large}$
CodeNN	8.58±0.15	$16.16 \pm 0.20$	$20.13 \pm 0.34$	$25.26 \pm 0.00$	$31.19 \pm 0.17$
Deepcom	$6.12 \pm 0.64$	$11.29 \pm 0.21$	$12.64 \pm 1.07$	$20.80 \pm 0.02$	$23.72 \pm 0.32$
Astattgru	$11.73 \pm 0.41$	$20.22 \pm 0.39$	$24.11 \pm 0.25$	$27.63 \pm 0.24$	$33.36 \pm 0.16$
Rencos	11.19±0.09	$19.75 \pm 0.10$	$25.03 \pm 0.02$	$25.82 \pm 0.00$	$33.95 \pm 0.03$
NCS	$11.80 \pm 0.94$	$23.25 \pm 0.13$	$30.12 \pm 0.27$	$30.69 \pm 0.12$	$40.73 \pm 0.16$
OOV Ratio	48.74%	35.38%	34.00%	57.56%	48.60%

Table 48: The result in different data splitting methods. Evaluated with BLEU-NCS.

Model	CSN <sub>Project-Medium</sub>	CSN <sub>Class-Medium</sub>	CSN <sub>Method-Medium</sub>	FCM <sub>Project-Large</sub>	$FCM_{Method-Large}$
CodeNN	$11.40 \pm 0.32$	$18.06 \pm 0.18$	$21.63 \pm 0.17$	$26.43 \pm 0.06$	31.96±0.19
Deepcom	$9.18\pm0.30$	$14.04 \pm 0.34$	$14.96 \pm 0.95$	$23.44 \pm 0.06$	$25.62 \pm 0.37$
Astattgru	$12.44 \pm 1.14$	$20.07 \pm 0.70$	$24.05 \pm 0.34$	$28.44 \pm 0.24$	$33.92 \pm 0.14$
Rencos	$14.76 \pm 0.08$	$22.62 \pm 0.05$	$27.64 \pm 0.00$	$28.23 \pm 0.00$	$35.91 \pm 0.02$
NCS	$14.68 \pm 0.62$	$25.77 \pm 0.17$	$32.55 \pm 0.27$	$31.18 \pm 0.34$	$40.71 \pm 0.28$

 ${\bf Table~49: The~result~in~different~data~splitting~methods.~Evaluated~with~BLEU-RC.}$ 

Model	CSN <sub>Project-Medium</sub>	CSN <sub>Class-Medium</sub>	${\rm CSN}_{\rm Method\text{-}Medium}$	FCM <sub>Project-Large</sub>	$FCM_{Method-Large}$
CodeNN	$1.56 \pm 0.11$	$6.83 \pm 0.08$	9.28±0.25	$7.28 \pm 0.14$	$12.64 \pm 0.25$
Deepcom	$0.99 \pm 0.25$	$4.31 \pm 0.22$	$5.79 \pm 0.78$	$4.96 \pm 0.04$	$6.97 \pm 0.36$
Astattgru	$5.14 \pm 4.05$	$8.55 \pm 0.41$	$12.32 \pm 0.22$	$9.94 \pm 0.06$	$15.80 \pm 0.09$
Rencos	$4.76 \pm 0.06$	$11.86 \pm 0.04$	$16.73 \pm 0.04$	$10.41 \pm 0.00$	$18.11 \pm 0.04$
NCS	$3.13 \pm 0.53$	$12.28 \pm 0.14$	$18.69 \pm 0.28$	$12.30 \pm 0.53$	$23.00 \pm 0.27$

Table 50: The result in different data splitting methods. Evaluated with Rouge.

Model	CSN <sub>Project-Medium</sub>	CSN <sub>Class-Medium</sub>	CSN <sub>Method-Medium</sub>	FCM <sub>Project-Large</sub>	FCM <sub>Method-Large</sub>
CodeNN	$16.01 \pm 0.38$	$24.77 \pm 0.24$	30.24±0.29	$34.98 \pm 0.05$	42.43±0.29
Deepcom	$10.86 \pm 0.23$	$17.10 \pm 0.30$	$19.00 \pm 0.87$	$28.13 \pm 0.05$	$32.04 \pm 0.61$
Astattgru	$22.20 \pm 1.76$	$29.61 \pm 0.81$	$34.21 \pm 0.64$	$38.59 \pm 0.22$	$44.94 \pm 0.15$
Rencos	$20.78 \pm 0.10$	$30.57 \pm 0.14$	$37.04 \pm 0.11$	$35.78 \pm 0.00$	$45.37 \pm 0.04$
NCS	$21.91 \pm 2.34$	$33.48 \pm 0.15$	$40.97 \pm 0.52$	40.61±0.19	$50.89 \pm 0.33$

 ${\bf Table~51:}~{\bf The~result~in~different~data~splitting~methods.~Evaluated~with~Meteor.}$ 

Model	CSN <sub>Project-Medium</sub>	CSN <sub>Class-Medium</sub>	CSN <sub>Method-Medium</sub>	FCM <sub>Project-Large</sub>	$FCM_{Method-Large}$
CodeNN	5.36±0.26	$9.74 \pm 0.14$	$11.53 \pm 0.12$	$14.84 \pm 0.21$	19.01±0.21
Deepcom	$3.72 \pm 0.35$	$6.76 \pm 0.23$	$7.47 \pm 0.25$	$10.42 \pm 0.01$	$12.48 \pm 0.28$
Astattgru	$8.68 \pm 1.01$	$11.47 \pm 0.31$	$13.20 \pm 0.04$	$17.72 \pm 0.10$	$21.40 \pm 0.07$
Rencos	$8.94 \pm 0.08$	$12.82 \pm 0.14$	$15.48 \pm 0.11$	$16.55 \pm 0.00$	$21.84 \pm 0.03$
NCS	$6.39 \pm 1.19$	$12.86 \pm 0.16$	$15.62 \pm 0.24$	$19.04 \pm 0.25$	$25.25 \pm 0.23$

Table 52: The result in different data splitting methods. Evaluated with Cider.

Model	CSN <sub>Project-Medium</sub>	CSN <sub>Class-Medium</sub>	${\rm CSN}_{\rm Method\text{-}Medium}$	FCM <sub>Project-Large</sub>	$FCM_{Method-Large}$
CodeNN	$0.32 \pm 0.02$	$0.90 \pm 0.01$	1.21±0.03	$1.36 \pm 0.01$	$1.95 \pm 0.03$
Deepcom	$0.15 \pm 0.03$	$0.53 \pm 0.01$	$0.68 \pm 0.09$	$0.76 \pm 0.01$	$1.01 \pm 0.04$
Astattgru	$0.36\pm0.19$	$1.06 \pm 0.06$	$1.46 \pm 0.03$	$1.78 \pm 0.02$	$2.41 \pm 0.02$
Rencos	$0.58 \pm 0.01$	$1.30 \pm 0.01$	$1.82 \pm 0.01$	$1.63 \pm 0.00$	$2.48 \pm 0.00$
NCS	$0.63 \pm 0.06$	$1.56 \pm 0.01$	$2.20{\pm}0.03$	$2.21 \pm 0.03$	$3.27 \pm 0.02$

Table 53: The result of different duplication ratios. Evaluated with BLEU-DM.

Model		(	duplicat	ion ratio	)	
	0.0	0.1	0.3	0.5	0.7	0.9
CodeNN	4.49	6.01	9.61	13.51	17.00	20.60
Deepcom	3.51	5.10	8.27	11.33	14.43	18.13
Astattgru	4.15	7.79	15.23	22.34	29.98	37.21
Rencos	21.52	29.25	44.60	59.94	75.26	90.64
NCS	10.14	17.10	26.04	35.22	43.77	53.51

Table 54: The result of different duplication ratios. Evaluated with BLEU-FC.

Model		(	duplicat	ion ratio	)	
1110 0001	0.0	0.1	0.3	0.5	0.7	0.9
CodeNN	5.31	6.69	9.59	12.42	15.21	17.73
Deepcom	4.03	5.20	7.51	9.65	11.59	14.08
Astattgru	6.39	9.85	15.96	22.14	28.13	33.73
Rencos	23.16	30.05	44.22	57.51	71.27	84.52
NCS	8.63	13.70	19.74	26.33	32.51	39.59

Table 55: The result of different duplication ratios. Evaluated with BLEU-CN.

Model		(	duplicat	ion ratio	)	
	0.0	0.1	0.3	0.5	0.7	0.9
CodeNN	13.71	15.63	19.49	23.79	27.57	31.59
Deepcom	11.26	13.21	17.03	20.84	24.56	28.88
Astattgru	12.28	15.64	22.5	29.22	36.19	43.0
Rencos	29.98	36.86	50.52	64.18	77.82	91.5
NCS	21.11	26.95	34.66	42.7	50.06	58.56

Table 56: The result of different duplication ratios. Evaluated with BLEU-NCS.

Model		(	duplicat	ion ratio	)	
	0.0	0.1	0.3	0.5	0.7	0.9
CodeNN	14.58	16.32	20.04	24.18	27.81	31.70
Deepcom	12.68	14.50	18.04	21.61	25.04	29.11
Astattgru	13.16	16.45	23.15	29.70	36.53	43.19
Rencos	30.52	37.34	50.89	64.44	77.97	91.54
NCS	21.29	27.11	34.64	42.57	49.77	58.13

# **REFERENCES**

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Table 57: The result of different duplication ratios. Evaluated with BLEU-RC.

Model		(	duplicat	ion ratio	)	
1110 401	0.0	0.1	0.3	0.5	0.7	0.9
CodeNN	4.49	6.01	9.62	13.51	17.00	20.60
Deepcom	3.51	5.10	8.27	11.33	14.43	18.13
Astattgru	4.15	7.79	15.23	22.34	29.98	37.21
Rencos	21.52	29.25	44.60	59.94	75.26	90.64
NCS	10.14	17.10	26.04	35.22	43.77	53.51

Table 58: The result of different duplication ratios. Evaluated with Rouge.

Model		(	duplicat	ion ratio	)	
	0.0	0.1	0.3	0.5	0.7	0.9
CodeNN	25.67	28.55	34.02	39.87	45.35	50.93
Deepcom	19.99	22.95	28.61	34.32	40.10	46.31
Astattgru	23.38	26.79	33.65	40.60	47.61	54.56
Rencos	42.57	48.23	59.51	70.78	82.05	93.34
NCS	36.34	41.34	48.78	56.38	63.66	71.63

Table 59: The result of different duplication ratios. Evaluated with Meteor.

Model		(	duplicat	ion ratio	)	
	0.0	0.1	0.3	0.5	0.7	0.9
CodeNN	8.89	10.23	12.73	15.33	17.87	20.22
Deepcom	6.87	8.23	10.90	13.58	16.30	19.16
Astattgru	8.44	10.01	13.28	16.86	20.53	24.12
Rencos	18.33	22.17	30.69	39.69	50.27	63.46
NCS	12.48	15.38	19.39	23.62	27.70	32.23

Table 60: The result of different duplication ratios. Evaluated with Cider.

Model		duplication ratio					
1.13461	0.0	0.1	0.3	0.5	0.7	0.9	
CodeNN	0.70	0.89	1.28	1.69	2.04	2.37	
Deepcom	0.51	0.70	1.07	1.44	1.81	2.22	
Astattgru	0.57	0.90	1.60	2.25	2.95	3.60	
Rencos	2.31	3.05	4.52	5.97	7.44	8.91	
NCS	1.43	2.04	2.84	3.67	4.41	5.25	

Table 61: Performance (BLEU-DM) in Different datasets with same data splitting way, corpus size, and duplication ratio.

Model	$TLC_{Dedup}$	$FCM_{Method\text{-}Small}$	$CSN_{Method\text{-}Small}$
CodeNN	$7.12 \pm 0.14$	$5.31 \pm 0.17$	$3.28 \pm 0.06$
Deepcom	$3.46 \pm 0.86$	$4.86 \pm 0.05$	$5.76 \pm 0.75$
Astattgru	$4.83 \pm 0.39$	$7.80 \pm 0.51$	$3.93 \pm 0.08$
Rencos	$21.48 \pm 0.03$	$9.48 \pm 0.17$	$5.61 \pm 0.07$
NCS	$18.42 \pm 0.10$	$8.98 \pm 0.09$	$5.16 \pm 0.13$

Table 62: Performance (BLEU-FC) in Different datasets with same data splitting way, corpus size, and duplication ratio.

Model	$TLC_{Dedup}$	$FCM_{Method\text{-}Small}$	$CSN_{Method\text{-}Small}$
CodeNN	$8.64 \pm 0.21$	$9.70 \pm 0.23$	$3.95 \pm 0.05$
Deepcom	$3.69 \pm 1.21$	$8.23 \pm 0.09$	$3.34 \pm 0.03$
Astattgru	$6.91 \pm 0.33$	$12.93 \pm 0.71$	$4.79 \pm 0.45$
Rencos	$23.56 \pm 0.13$	$14.86 \pm 0.15$	$7.35 \pm 0.52$
NCS	$18.33 \pm 0.34$	$13.66 \pm 0.40$	$2.61 \pm 0.14$

Table 63: Performance (BLEU-DC) in Different datasets with same data splitting way, corpus size, and duplication ratio.

Model	$TLC_{Dedup}$	$FCM_{Method\text{-}Small}$	CSN <sub>Method-Small</sub>
CodeNN	9.69±0.10	10.37±0.17	5.20±0.01
Deepcom	$5.65 \pm 1.14$	$8.99 \pm 0.06$	$7.57 \pm 0.74$
Astattgru	$7.07 \pm 0.50$	$12.86 \pm 0.64$	$5.89 \pm 0.12$
Rencos	$24.36 \pm 0.00$	$14.24 \pm 0.12$	$7.36 \pm 0.08$
NCS	$21.67 \pm 0.09$	$14.70 \pm 0.19$	$9.07 \pm 0.20$

Table 64: Performance(BLEU-CN) in Different dataset with same split way, corpus size, and duplication ratio.

Model	$TLC_{Dedup}$	$FCM_{Method\text{-}Small}$	$CSN_{Method-Small}$
CodeNN	15.59±0.05	$22.85 \pm 0.12$	$9.38 \pm 0.14$
Deepcom	$10.77 \pm 1.73$	$20.49 \pm 0.16$	$12.30\pm0.64$
Astattgru	$12.49 \pm 0.59$	$23.94 \pm 0.67$	$11.38 \pm 0.42$
Rencos	$29.85 \pm 0.03$	$24.02 \pm 0.03$	$11.73 \pm 0.16$
NCS	$27.38 \pm 0.14$	$28.17 \pm 0.00$	$12.67 \pm 0.07$

Table 65: Performance (BLEU-NCS) in Different datasets with same data splitting way, corpus size, and duplication ratio.

Model	$TLC_{Dedup}$	$FCM_{Method\text{-}Small}$	CSN <sub>Method-Small</sub>
CodeNN	16.69±0.07	24.45±0.09	$12.57 \pm 0.13$
Deepcom	$12.06 \pm 1.88$	$23.18 \pm 0.06$	$14.49 \pm 0.23$
Astattgru	$13.36 \pm 0.65$	$25.53 \pm 0.64$	$12.99 \pm 0.13$
Rencos	$30.92 \pm 0.03$	$26.83 \pm 0.08$	$15.04 \pm 0.16$
NCS	$28.96 \pm 0.10$	$28.29 \pm 0.27$	$16.57 \pm 0.16$

Table 66: Performance (BLEU-RC) in Different datasets with same data splitting way, corpus size, and duplication ratio.

Model	$TLC_{Dedup}$	FCM <sub>Method-Small</sub>	CSN <sub>Method-Small</sub>
CodeNN	$7.12 \pm 0.14$	$5.32 \pm 0.17$	$3.28 \pm 0.06$
Deepcom	$3.46 \pm 0.86$	$4.87 \pm 0.05$	$5.76 \pm 0.75$
Astattgru	$4.83 \pm 0.39$	$7.82 \pm 0.51$	$3.93 \pm 0.08$
Rencos	$21.48 \pm 0.03$	$9.49 \pm 0.17$	$5.62 \pm 0.07$
NCS	$18.42 \pm 0.10$	$8.98 \pm 0.09$	$5.16 \pm 0.13$

Table 67: Performance (Rouge) in Different datasets with same data splitting way, corpus size, and duplication ratio.

Model	$TLC_{Dedup}$	$FCM_{Method\text{-}Small}$	$CSN_{Method\text{-}Small}$
CodeNN	26.64±0.35	$32.17 \pm 0.22$	16.53±0.15
Deepcom	$18.92 \pm 3.22$	$27.68 \pm 0.15$	$18.64 \pm 0.35$
Astattgru	$22.77 \pm 0.98$	$34.08 \pm 1.02$	$18.37 \pm 0.11$
Rencos	$42.88 \pm 0.02$	$33.56 \pm 0.03$	$19.87 \pm 0.20$
NCS	$41.76 \pm 0.34$	$37.83 \pm 0.86$	$21.22 \pm 0.29$

Table 68: Performance (Meteor) in Different datasets with same data splitting way, corpus size, and duplication ratio.

Model	$TLC_{Dedup}$	$FCM_{Method\text{-}Small}$	$CSN_{Method\text{-}Small}$
CodeNN	11.99±0.18	12.93±0.20	5.95±0.04
Deepcom	$5.79 \pm 1.77$	$9.97 \pm 0.01$	$7.47 \pm 0.25$
Astattgru	$8.86 \pm 0.47$	$14.87 \pm 0.64$	$6.37 \pm 0.20$
Rencos	$22.97 \pm 0.16$	$15.00 \pm 0.03$	$7.67 \pm 0.21$
NCS	$23.12 \pm 0.22$	$16.71 \pm 0.33$	$6.97 \pm 0.07$

Table 69: Performance (Cider) in Different datasets with same data splitting way, corpus size, and duplication ratio.

Model	$TLC_{Dedup}$	FCM <sub>Method-Small</sub>	CSN <sub>Method-Small</sub>
CodeNN	0.99±0.01	$0.99 \pm 0.02$	$0.44 {\pm} 0.01$
Deepcom	$0.57 \pm 0.11$	$0.69 \pm 0.02$	$0.67 \pm 0.08$
Astattgru	$0.61 \pm 0.06$	$1.25 \pm 0.10$	$0.45 \pm 0.00$
Rencos	$2.30 \pm 0.00$	$1.37 \pm 0.01$	$0.64 \pm 0.01$
NCS	$2.13 \pm 0.01$	$1.82 \pm 0.01$	$0.75 \pm 0.02$