

baselines

2019年3月13日 14:29

		baselines	evaluations	automatic evaluations																																																																																					
planning	baidu 16 COLING																																																																																								
polishing (I, Poet)	Yan 16		Perplexity BLEU human	<table border="1"> <thead> <tr> <th rowspan="2">Algo.</th> <th colspan="3">5-Character</th> <th colspan="3">7-Character</th> </tr> <tr> <th>PPL</th> <th>BLEU</th> <th>Human</th> <th>PPL</th> <th>BLEU</th> <th>Human</th> </tr> </thead> <tbody> <tr> <td>Random</td> <td>—</td> <td>0.002</td> <td>0.259</td> <td>—</td> <td>0.051</td> <td>0.135</td> </tr> <tr> <td>SMT</td> <td>126</td> <td>0.051</td> <td>1.943</td> <td>134</td> <td>0.144</td> <td>1.957</td> </tr> <tr> <td>SUM</td> <td>149</td> <td>0.035</td> <td>2.219</td> <td>131</td> <td>0.128</td> <td>2.013</td> </tr> <tr> <td>RNNPG</td> <td>103</td> <td>0.053</td> <td>1.964</td> <td>119</td> <td>0.163</td> <td>2.205</td> </tr> <tr> <td>LSTM-RNN</td> <td>123</td> <td>0.048</td> <td>1.762</td> <td>136</td> <td>0.159</td> <td>1.633</td> </tr> <tr> <td>iPoet</td> <td>91</td> <td>0.088</td> <td>2.352</td> <td>96</td> <td>0.185</td> <td>2.568</td> </tr> </tbody> </table>							Algo.	5-Character			7-Character			PPL	BLEU	Human	PPL	BLEU	Human	Random	—	0.002	0.259	—	0.051	0.135	SMT	126	0.051	1.943	134	0.144	1.957	SUM	149	0.035	2.219	131	0.128	2.013	RNNPG	103	0.053	1.964	119	0.163	2.205	LSTM-RNN	123	0.048	1.762	136	0.159	1.633	iPoet	91	0.088	2.352	96	0.185	2.568	Perplexity (PPL). For most of the language generation research, language perplexity is a sanity check. Our first set of experiments involved intrinsic evaluation of the "perplexity" evaluation for the generated poems. Perplexity is actually an entropy based evaluation. In this sense, the lower perplexity for the poems generated, the better performance in purity for the generations, and the poems are likely to be good ones.																							
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