

A Rules and Examples

Table 8: Number of rules in \mathbb{G}^{KB}

KB	PPDB	SICK	WORDNET
#Rules	6,977,679	12,511	~116,000
Examples	because of \Rightarrow due to, wish \Rightarrow would like	woods \Rightarrow wooden area, kid \nRightarrow woman	car \Rightarrow cabin car, hate \nRightarrow love

Table 8 shows the number of rules and additional examples for \mathbb{G}^{KB} .

B Training data sizes

Figure 3 shows training (dotted) accuracies on sub-sampled training datasets and testing (solid) accuracies on original test dataset X_{test} of \mathbb{D} over different sub-sampling percentages of the training set. Since SciTail (27K) is much smaller than SNLI (570K), SciTail fluctuates a lot at smaller sub-samples while SNLI converges with just 50% of the examples.

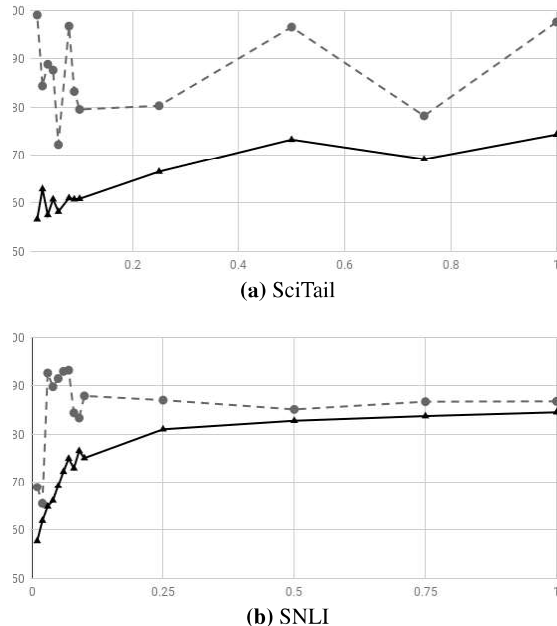


Figure 3: \mathbb{D} for SciTail and SNLI.

C Effectiveness of Z/X Ratio, α

Figure 4 shows train/test accuracies with different balancing ratio between z and x . The dotted line is training accuracies, the solid black horizontal line is testing accuracy of \mathbb{D} . The solid red shows

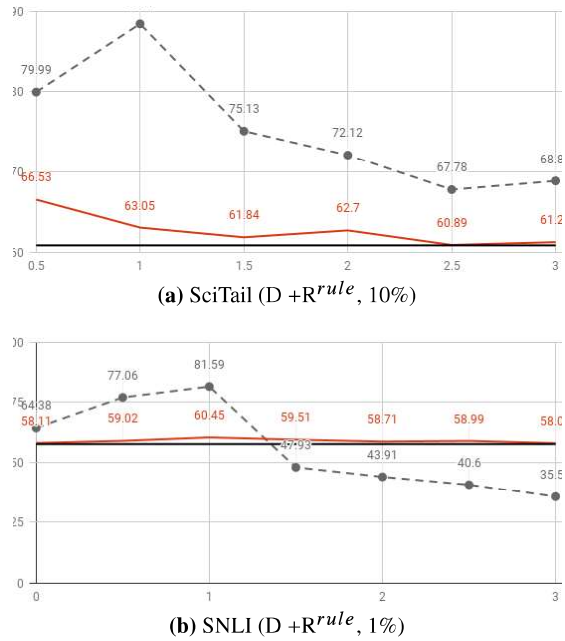


Figure 4: Effect of balancing ratio between z and x .

test accuracies with different balancing ratio, α (x-axis) from 0.5, 1.0, ... 3.0 from $|z| = \alpha * |x|$ where $|x|$ is fixed as batch size. The generated examples z are useful up to a point, but the performance quickly degrades for $\alpha > 1.0$ as they overwhelm the original dataset x .

D Retrofitting Experiment

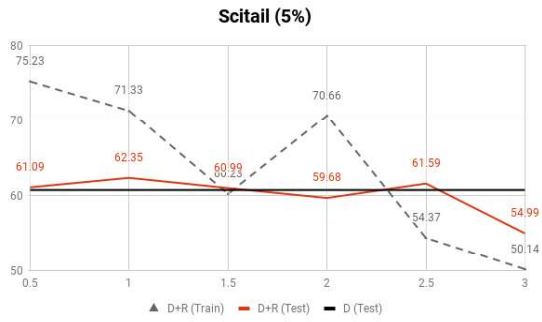
Table 9 shows the grid search results of retrofitting vectors (Faruqui et al., 2015) with different lexical resources. To obtain the strongest baseline, we choose the best performing vectors for each sub-sample ratio and each dataset. Usually, PPDB and WordNet are two most useful resources for both SNLI and SciTail.

E In-Depth Analysis: D+R

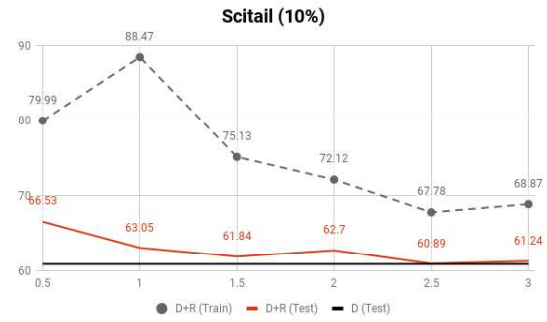
Table 5 and Table 6 show more in-depth analysis with different sub-sampling ratio on SNLI and SciTail. The dotted line is training accuracy, and the solid red ($\mathbb{D} + \mathbb{G}^{\text{rule}}$) and solid black (\mathbb{D}) shows testing accuracies.

Table 9: Results of the word vectors retrofitted on different lexicons on each dataset. We pick the best vectors for each task and sub-sampling ratio.

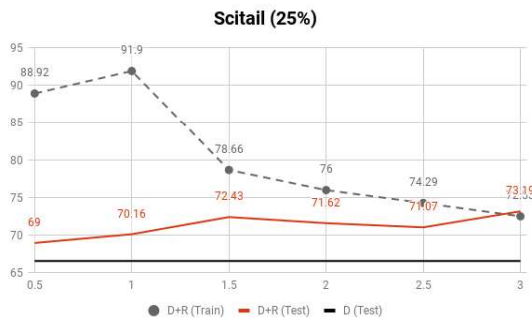
ratio	Lexicon	SNLI	SciTail
1%	framenet	56.15	60.89
1%	ppdb	57.04	62.5
1%	wordnet	55.58	62.2
1%	all	56.81	61.14
10%	framenet	72.75	67.99
10%	ppdb	72.88	54.74
10%	wordnet	73.27	67.29
10%	all	73.45	66.43
50%	framenet	80.95	66.08
50%	ppdb	81.14	67.24
50%	wordnet	80.62	69.05
50%	all	81.18	68.4
100%	framenet	83.66	70.06
100%	ppdb	84.14	70.16
100%	wordnet	83.91	72.63
100%	all	83.68	71.12



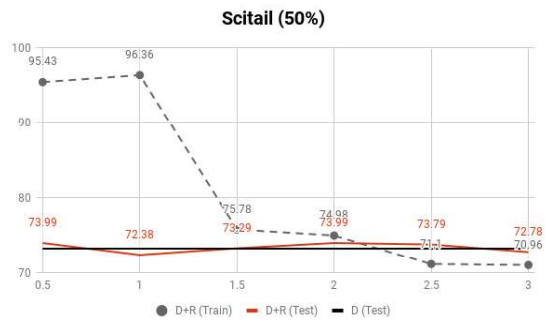
(a) D+R (5%)



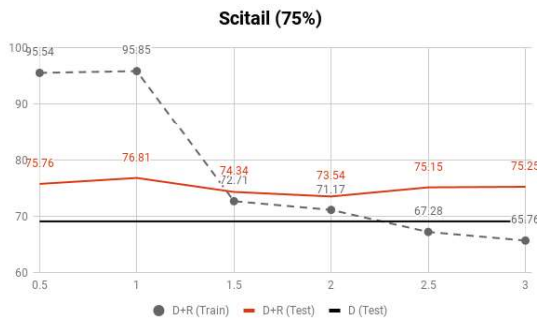
(b) D+R (10%)



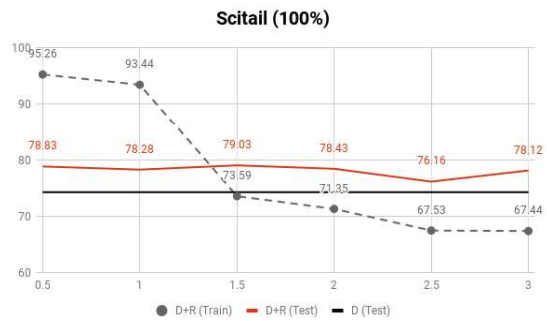
(c) D+R (25%)



(d) D+R (50%)



(e) D+R (75%)



(f) D+R (100%)

Figure 5: $\mathbb{D} + G^{\text{rule}}$ with different ratio for SciTail.

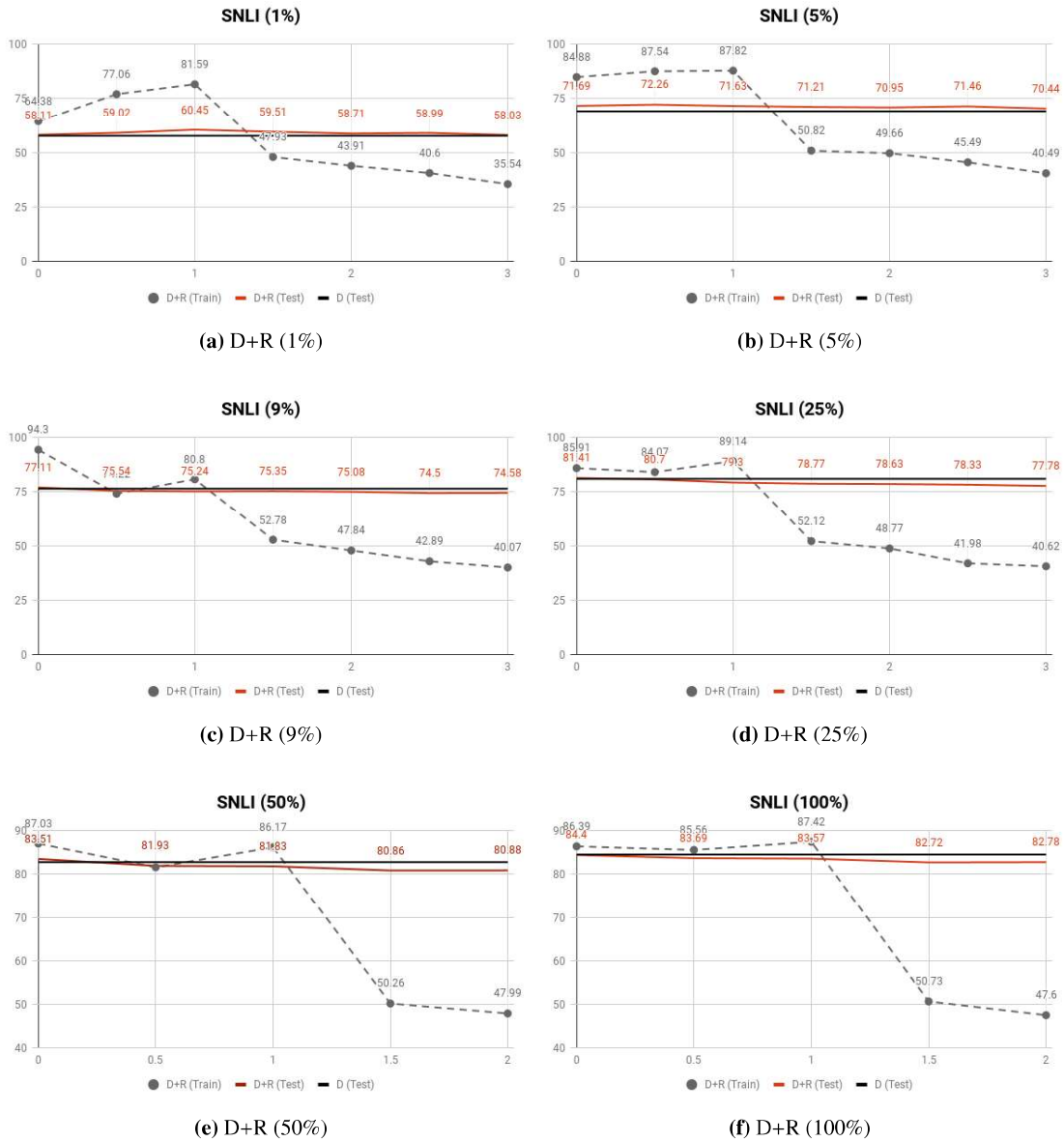


Figure 6: $\mathbb{D} + \mathbb{G}^{\text{rule}}$ with different ratio for SNLI.