Distill or **Annotate?** Cost-Efficient Fine-Tuning of Compact Models



Junmo Kang



Wei Xu



Alan Ritter







Q. Given a fixed budget, how to build a compact model in a cost-efficient way?



Introduction







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Strategy 1: annotate more data to directly train a small model



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Cost Estimations

Task & Annotation Cost

Dataset	Task	\$ per Label
WLP	Named Entity Recognition	\$0.26
S TANCEOSAURUS	Stance Classification	\$0.364
FEVER	Fact Verification	\$0.129
MULTIPITID	Paraphrase Identification	\$0.2
MULTIPITGEN	Paraphrase Generation	\$0.371
Natural Questions	Question Answering	\$0.129





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Computational Cost

\$1.875 per 1 GPU hour (est. based on A100 in Google Cloud Platform)





Main Results





Evaluation

Main Results





Evaluation

Main Results





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Main Results



STANCEOSAURUS (F1)

Starting #Data: 5K (\$1820)



FEVER (Accuracy)

Starting #Data: 5K (\$645)



MULTIPITId (Accuracy) Starting #Data: 5K (\$1000) 80.6 80.5 81.1 81.9



MULTIPITGen (BERT-iBLEU)



NATURAL QUESTIONS (F1)

Starting #Data: 10K (\$1290)



T5-Small (Ann.) T5-XXL => T5-Small (Dist.)



Main Results

WLP (F1) Starting #Data: 1K (\$260)



STANCEOSAURUS (F1)

Starting #Data: 1K (\$364)



FEVER (Accuracy)

Starting #Data: 1K (\$129)



MULTIPITId (Accuracy) Starting #Data: 1K (\$200)



MULTIPITGen (BERT-iBLEU)



NATURAL QUESTIONS (F1) Starting #Data: 1K (\$129)



T5-Small (Ann.) T5-XXL => T5-Small (Dist.)



Evaluation













Evaluation





Evaluation





Evaluation



Evaluation







GPT-3.5 as an Annotator

Knowledge Distillation with APIs

(GPT-3 as cheaper annotators*)



*Wang et al., 2022, Want To Reduce Labeling Cost? GPT-3 Can Help?



Evaluation

GPT-3.5 as an Annotator



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Dist.: \$161 (81.0 F1) - max

Ann. : \$1,980 (81.0 F1)

\$17,443 (87.5 F1) - max



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 could be cost-efficient compared to
 humans, but still limited



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 More details and analyses in the paper, such as different sizes of large & compact models