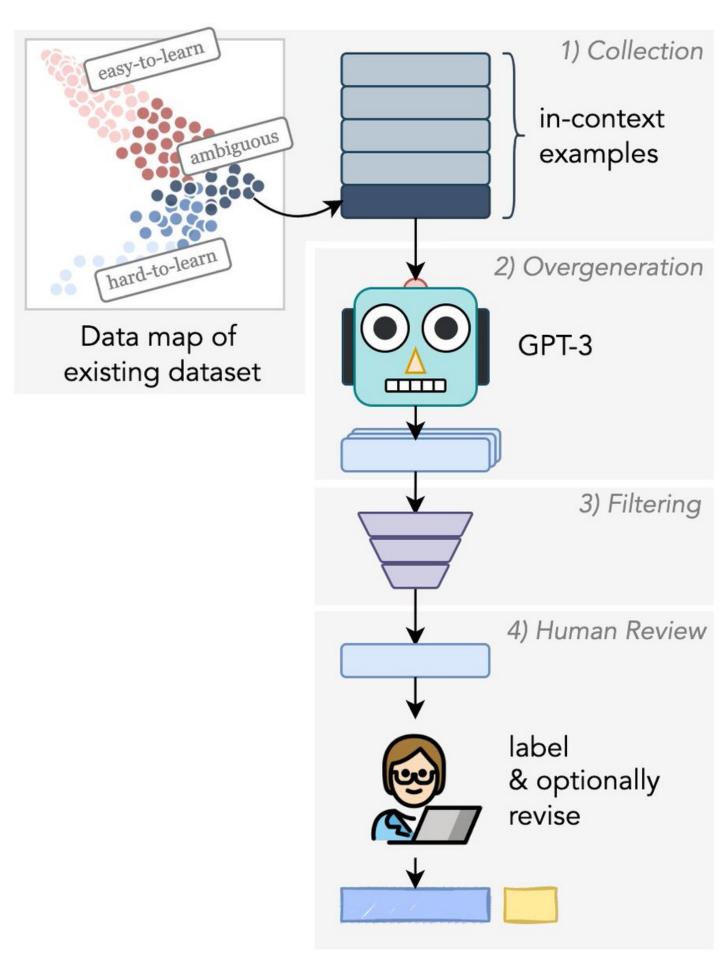
Twitter Thread by Alisa Liu

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We introduce a new paradigm for dataset creation based on human ■■■ and machine ■ collaboration, which brings together the generative strength of LMs and the evaluative strength of humans. And we collect ■ WaNLI, a dataset of 108K NLI examples! ■

Paper: https://t.co/IUXcm9wlh2



Our pipeline starts with an existing dataset (MNLI), and uses data maps ■ to automatically identify pockets of examples that demonstrate challenging ■ reasoning patterns relative to a trained model. Then we use GPT-3 to generate new examples likely to have the same pattern. 2/

Seed MNLI example	Generated WANLI Example	Label & Reasoning
P: 5 percent probability that each part will be defect free. H: Each part has a 95 percent chance of having a defect.	P: <i>1 percent</i> of the seats were vacant. H: <i>99 percent</i> of the seats were occupied.	Entailment Set complements
P: To the <i>south</i> , in the Sea of Marmara, lie the woods and beaches of the Princes' Islands. H: In the <i>north</i> is the Sea of Marmara where there are mountains to climb.	P: From the park's <i>southern entrance</i> , follow the avenue <i>south</i> to the Hotel de Ville. H: From the park's <i>northern entrance</i> , follow the avenue <i>north</i> to the Hotel de Ville.	Contradiction Reversing cardinal directions
P: To build a worldclass finance organization and help achieve better business outcomes, each of the organizations we examined <i>set an agenda for transforming</i> the finance organization by defining a shared vision -i.e. H: <i>The transformation was a disaster</i> and the entire organization had to be scrapped.	P: In order to help improve customer service, <i>I suggested</i> that they send a representative to our office to discuss our concerns. H: The representative sent to our office did not solve our problems and we lost a lot of business.	Neutral Intended goals may not actualize
P: Salinger <i>wrote</i> similar letters <i>to</i> other young female writers. H: Other young female writers <i>received</i> similar letters <i>from</i> Salinger as well.	P: The three schools <i>have</i> a number of students who are from families with no history of financial difficulties. H: Families with no history of financial difficulties <i>send</i> their children to the three schools.	Entailment Substituting a verb with a different subcategorization fram

Table 1: Seed MNLI examples, and corresponding WANLI examples which were fully generated by GPT-3. P stands for premise, H for hypothesis. The seed example is "ambiguous" according to the definitions of Swayamdipta et al. (2020), discussed in §2. The remaining in-context examples (shown in Appendix C) share the same pattern and are found using distance in [CLS] embeddings of a trained task model. The reasoning is a short description of the pattern we observe from the group, and which is successfully repeated in the generated example.

Next we propose a new metric, also inspired by data maps, to automatically filter generations for those most likely to aid model learning. Finally, we validate ■ the generated examples through crowdworkers, who assign a gold label ■ and (optionally) revise for quality △■. 3/

Remarkably, replacing MNLI with WaNLI (which is 4x smaller) for training improves performance■ on seven OOD test sets■, including by 11% on HANS and 9% on ANLI. Under a data augmentation setting, combining MNLI with WaNLI is more effective than using other augmentation sets. 4/

Our method addresses limitations of crowdsourcing, where workers may resort to repetitive writing strategies ■, and leverages the great progress in text generation ■. We get the best of both worlds: ■'s ability to produce diverse examples, and ■■■'s ability to evaluate them. 5/

We hope our work demonstrates the promise of leveraging LMs in a controlled way to aid the dataset creation process, and encourage the community to think of dataset curation as an AI challenge itself ■. Co-authored with @swabhz @nlpnoah @YejinChoinka ■ 6/6