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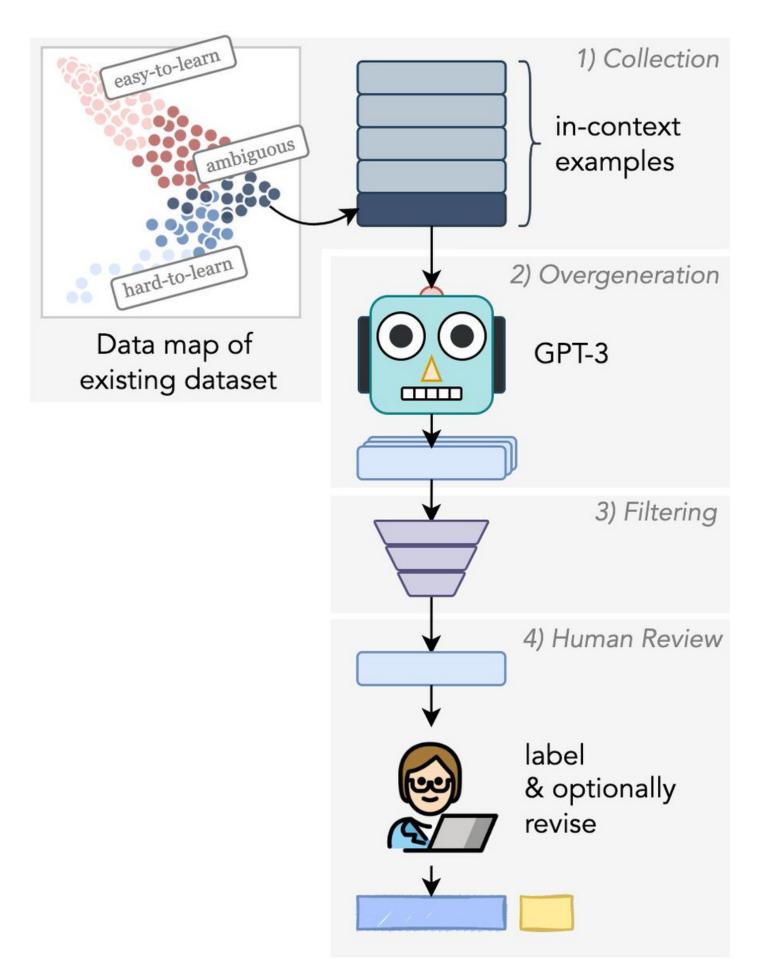
## Twitter Thread by Alisa Liu

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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/IUXcm9wIh2



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
<ul><li>P: <i>5 percent</i> probability that each part will be defect free.</li><li>H: Each part has a <i>95 percent</i> chance of having a defect.</li></ul>	<ul><li>P: 1 percent of the seats were vacant.</li><li>H: 99 percent of the seats were occupied.</li></ul>	Entailment Set complements
<ul><li>P: To the <i>south</i>, in the Sea of Marmara, lie the woods and beaches of the Princes' Islands.</li><li>H: In the <i>north</i> is the Sea of Marmara where there are mountains to climb.</li></ul>	<ul> <li>P: From the park's <i>southern entrance</i>, follow the avenue <i>south</i> to the Hotel de Ville.</li> <li>H: From the park's <i>northern entrance</i>, follow the avenue <i>north</i> to the Hotel de Ville.</li> </ul>	Contradiction Reversing cardinal directions
<ul> <li>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.</li> <li>H: <i>The transformation was a disaster</i> and the entire organization had to be scrapped.</li> </ul>	<ul> <li>P: In order to help improve customer service, <i>I suggested that they send a representative</i> to our office to discuss our concerns.</li> <li>H: <i>The representative</i> sent to our office <i>did not solve our problems</i> and we lost a lot of business.</li> </ul>	<i>Neutral</i> 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.	<ul> <li>P: The three schools <i>have</i> a number of students who are from families with no history of financial difficulties.</li> <li>H: Families with no history of financial difficulties <i>send</i> their children to the three schools.</li> </ul>	<i>Entailment</i> 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  $\blacksquare$ , and leverages the great progress in text generation  $\blacksquare$ . We get the best of both worlds:  $\blacksquare$ 's ability to produce diverse examples, and  $\blacksquare\blacksquare\blacksquare$ '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  $\blacksquare$ . Co-authored with <u>@swabhz</u> <u>@nlpnoah</u> <u>@YejinChoinka</u>  $\blacksquare$  6/6