AutoML: Introduction
Risks of AutoML

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1. Users apply AutoML without understanding anything.
   - Users might wonder why (Auto-)ML does not perform well after they passed in poor data.

2. Users trust AutoML too much.
   - humans might not use human reasoning skills and do not second guess machine decisions

③ We enable non-ML experts to use ML
without knowing the risks and consequences of ML.
  ▸ E.g., bias in data and trained models

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   ▶ biased and unfair models due to lack of understanding ethical practices and use of features such as gender and race for predicting outcomes

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   - ⇝ Future research topics (as of March 2020)

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3. Can we automatically find ML systems that are less prone to such risks?
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