# A human-labeled Landsat-8 contrails dataset

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<sup>\*</sup> Equal contribution

<sup>&</sup>lt;sup>1</sup> Google

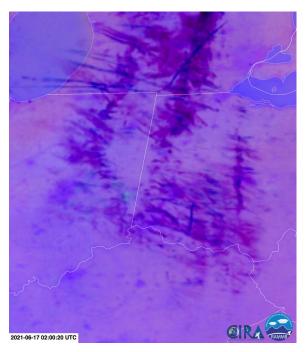
<sup>&</sup>lt;sup>2</sup> Laboratory for Aviation and the Environment, MIT

# Contrails (condensation trails) are anthropogenic clouds



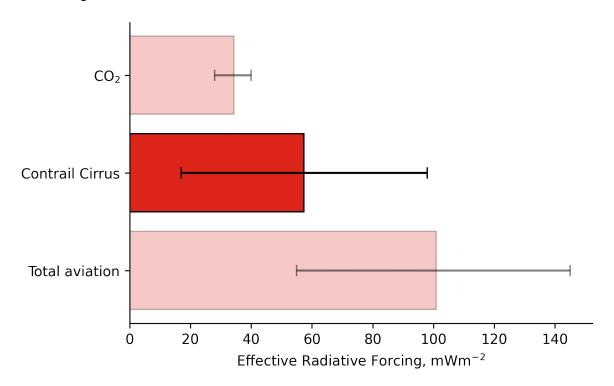


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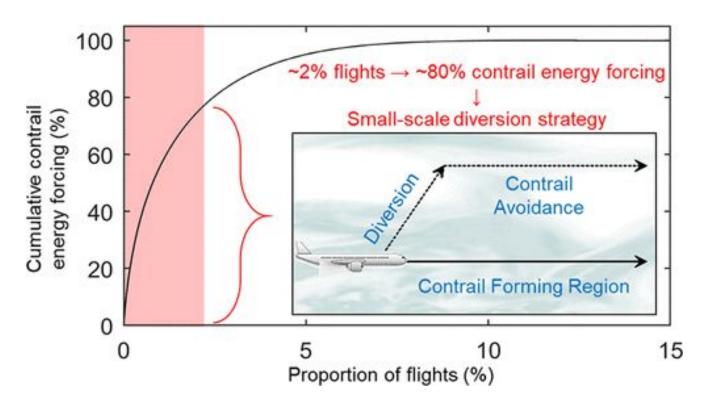


GOES-16 data rendered by https://rammb-slider.cira.colostate.edu/

## Contrails likely cause more than half of aviation's warming



## Small flight diversions may mitigate most of this warming



#### Satellite Contrail Detection is foundational

High accuracy is needed for:

- Evaluating contrail prediction models
- Tracking contrails as they evolve (thus measuring net warming impact)
- Verifying flight diversion effectiveness

#### Contrail remote sensing is ripe for ML - but labels are needed

Mannstein et al. 1999 has been state of the art in Satellite Contrail Detection.

Hand engineered algorithm with ~10 tunable parameters

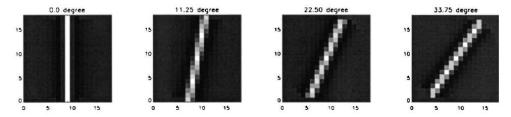
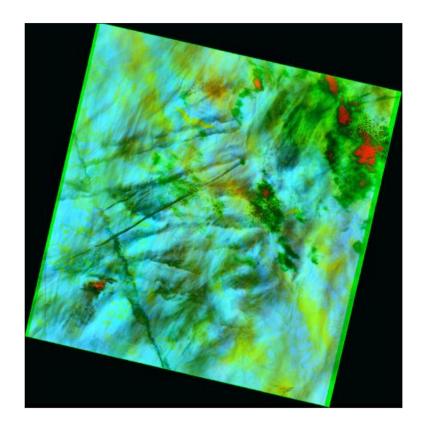


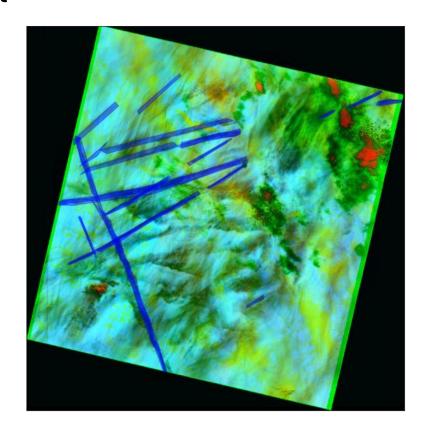
Figure 6. Example of the first four kernels for line filtering.

Only aware of one deep learning publication: Kulik, 2019/Meijer et al 2021 (under review)

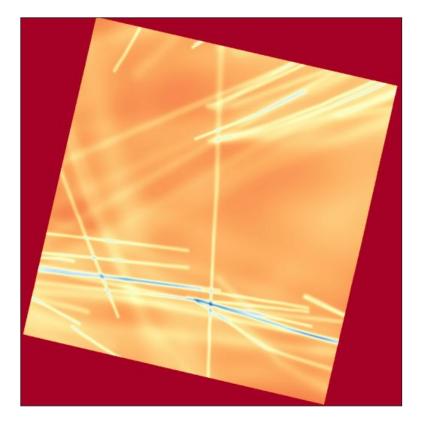
They created a labeled dataset on GOES-16 imagery

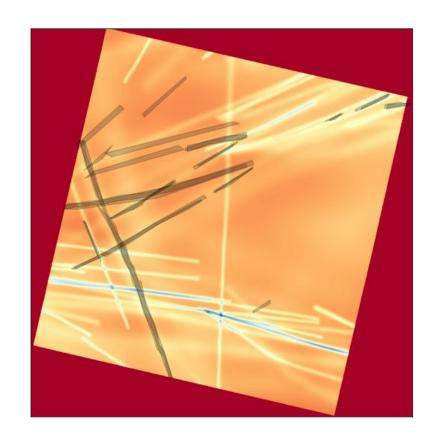
#### Our Landsat-8 contrails dataset





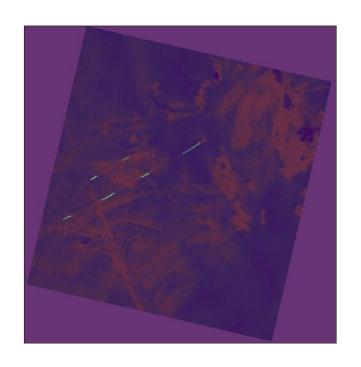
# Advected flights feature

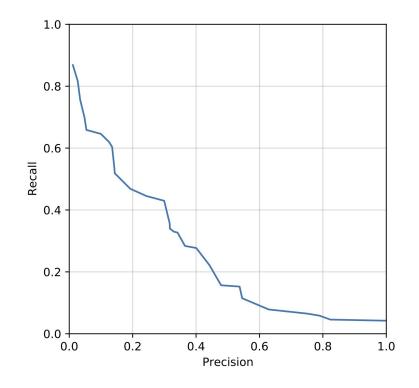




## Baseline performance for contrail detection in our dataset

#### Mannstein et al 1999 algorithm





#### Other possible uses for the dataset

- Validating high confidence subsets of GOES-16 detections
- training data for super-resolution models focused on contrail remote sensing
- analyzing how frequently contrails exist but are invisible in GOES-16 imagery
- evaluating contrail prediction models