



Is Your Face Fake?

Social Impacts of Algorithmic 'Fake' Determination (DFDC Risk-a-thon & CVPR medial forensics workshop 2020)

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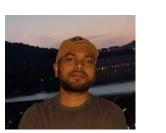


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Facebook AI (Sponsor)



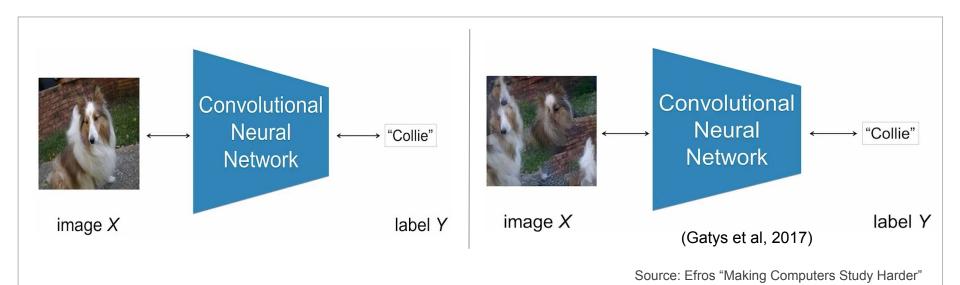
The Big Challenge: Deepfake Detection







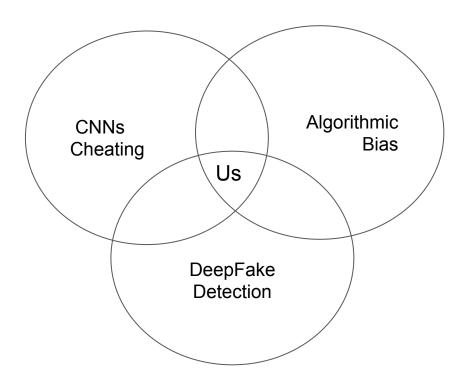
Deep CNNs + large & diverse datasets → 'Fake' & 'Real' classes





Our Direction





Cheating detection of blending artifacts; which affect already people marginalized by face



#1: Blending Artifacts → Colour Gradients







Original

Fake





More clear examples

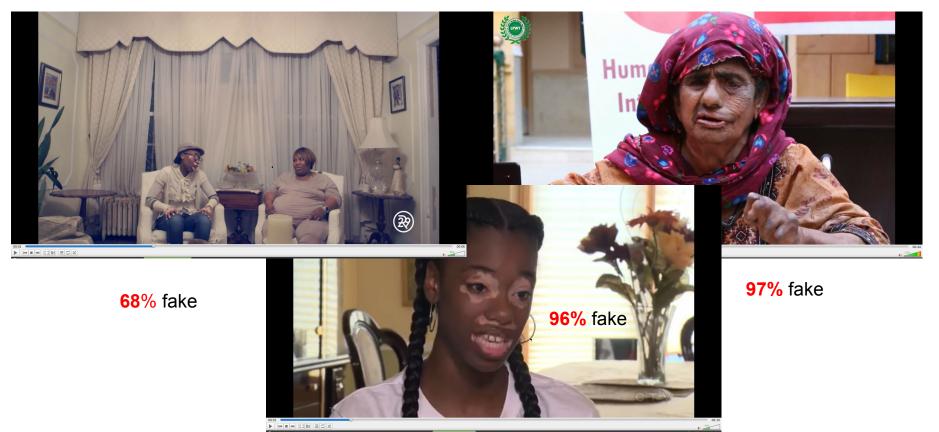


Vitiligo Victims



People with Leprosy & Vitiligo



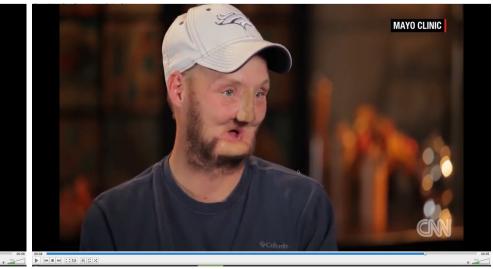




Burn victims







81% fake **96%** fake



#2: Blending Artifacts → Jaggedness









Original Fake Face Tattoos



(c) Face Tattoos





62% Fake **79%** Fake



#3: Blending Artifacts → Smooth Faces





Original



Fake





Babies







Results-at-a-Glance and Take Aways



Quantitative Results

- Babies (Rate: 14/20 babies classified fake)
- Burn victims (Rate: 2/5 people with burns classified fake)
- Leprosy/vitiligo victims (Rate: 1/5 and 2/5 people classified fake)
- Face Tattoos (Rate: 2/10 rappers classified fake)

No Consistent Patterns

- Need detailed further to identify any systemic pattern currently
- The models are brittle to a variety of artifacts even on high-res videos, with no changes

Need to prevent a lot more varieties of o.o.d false positives...

- Genuine editing: Instagram filters
- Heavy makeup, piercings, etc. should not be classified 'fake'











Thank You!