



Trusting Al systems

- Any practical Al system in production needs to be:
- Fair
 - Not allowing for any bias or discrimination
- Robust
 - Not able to be manipulated from the outside
- Explainable
 - Able to understand the internal decision process
- Need for Al governance and responsible Al
 - Technical solutions exist, but at some costs (e.g., slower execution)



Fairness

No discrimination against minorities or bias in decisions

- Bias is often present in data and transferred into models
 - Toxic effects of reinforcing existing unhealthy stereotypes
- Some recent examples
 - Facial recognition worked better for light-skinned males (Buolamwini)
 - Man is to computer programmer as women is to homemaker? (Bolukbasi)
 - Amazon's hiring tool discriminated against women (<u>Reuters</u>)



Robustness

- Not able to be manipulated by a third party via adversarial attacks
 - Deliberately force to make a wrong prediction and trying to fool the Al
- Make the system do something else than it is intended to do:
 - Stickers on stop sign confuse the Al
 - Patch that tricks AI into thinking a banana is a toaster
 - Glasses make facial recognition AI think you're actress Milla Jovovich
- Adversarial use of Al
 - Obama Deep Fake video



Explainability

- Understand why a specific decision is made
 - User has the "right to an explanation" (GDPR)
 - Especially important for high-stakes decisions with a big impact on lives
- Wolf vs husky experiment (<u>Ribeiro et al.</u>)
 - Snow in the background? → Husky
- Two options to guarantee explainability
 - Tranparent models
 - Ex-post interpretation techniques of black box models (many exist)