







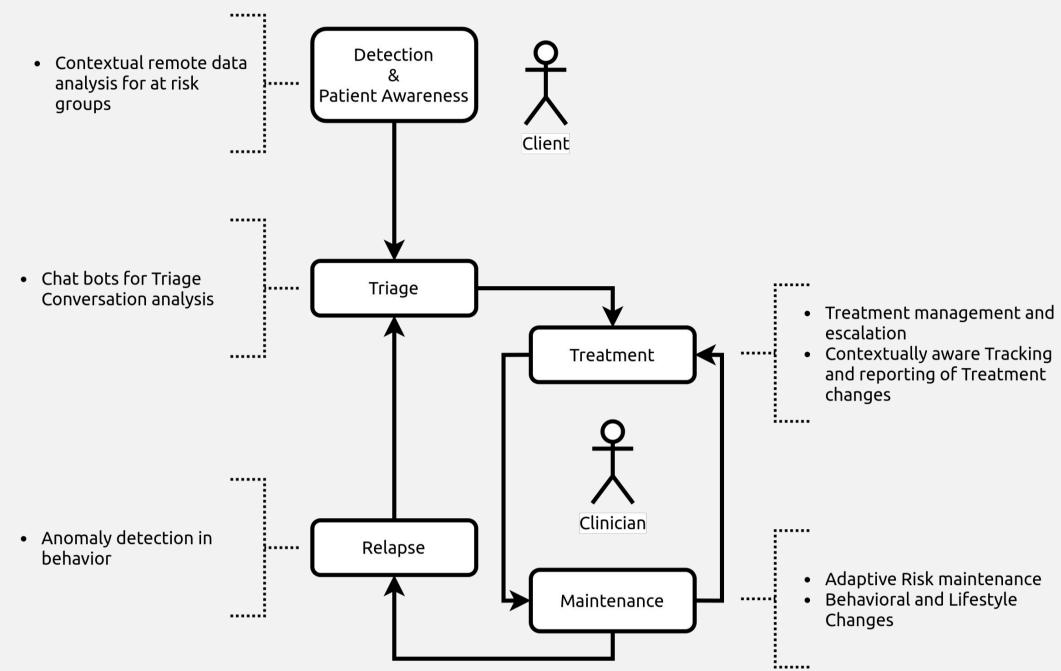
## Designing Fair Machine Learning and Artificial Intelligence in Digital Mental Health

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Digital Mental Health is in a strong position to take advantage of the benefits of Machine learning and Artificial intelligence. Large quantities of longitudinal data are generated during digital health interventions. As the field beings to assess this data using the tools of advanced statistical analysis there are many areas in which client support can benefit.

Healthcare also contains a logical succession of potential implementations that lead from low-risk areas, such as the timing of meditation reminders, to areas requiring high or perfect accuracy such as medication reminders. The practices, models, and safeguards developed for one logically support the research of the next.

Mental health services are more in demand than ever and the data used by people in their day to day life, generated by IOT devices, could become an important part of how a modern health care system supports patients. These digital services should be met with a clear and transparent analysis of their safety and efficiency. Key to this is analysis of **Fairness** in their decisions.



| Some Fairness Definitions         |                                   |                                   |                              |
|-----------------------------------|-----------------------------------|-----------------------------------|------------------------------|
| Statistical Parity                | Equalised Odds                    | Well-Calibration                  | Fairness through awareness   |
| Conditional Statistical Parity    | Fairness through unawareness      | Balance for Positive Class        | Counterfactual Fairness      |
| Predictive Parity                 | Overall Accuracy Equality         | Balance for Negative Class        | No unresolved discrimination |
| Test-fairness or calibration      | Treatment Equality                | Causal discrimination             | No proxy discrimination      |
| False Negative Error Rate Balance | False Positive Error Rate Balance | Conditional Use Accuracy Equality | Fair Inference               |

Verma, S., & Rubin, J. (2018). Fairness definitions explained. Proceedings - International Conference on Software Engineering, 1–7. https://doi.org/10.1145/3194770.3194776

## Q. With ~27+ definitions of Fairness being incorporated into ethical standards how does a researcher build AI/ML into Medical Applications?

## A. Very Carefully

**Fairness** is the term used for practices of ensuring that no person or group of people is treated worse than any other. This, along with Transparency and Accountability, is the underpinning for how most researchers approach safe and ethical Machine Learning.

Many common definitions of **Fairness** focus on the statistical parity between different demographic groups. These parity metrics use identified variables on which there is a risk of discrimination (examples being gender, ethnicity, sexual orientation). These definitions are appropriate in some areas of digital mental health but can be absurd to consider in others. Imagine a Healthcare system focused on postpartum mental health that looks for parity between genders.

**Fairness** needs to be treated as more than a statistical or Machine Learning problem but as a sociological and designed issue. There are tools in the Human-Computer Interaction field that can potentially be used in the resolution of issues like this, however, work needs to be done to find their role and the language we need when talking about them.









