

A Practical Overview of Safety Concerns and Mitigation Methods for Visual Deep Learning Algorithms

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Objectives

- Categorizing faults and underlying causes in a visual deep learning algorithm
- Providing a practical and complete list of safety concerns in each stage
- Listing potential state-of-the-art mitigation methods to deal with the concerns
- Discussing the effectiveness of the mitigation methods

Safety Concerns and Mitigation Methods

- **Incomplete Dataset**
- **Inadequate Dataset**
- **Insufficient / Noise Dataset**
- **Ill-Matched Architecture**
- **Unfitting Metrics**
- **Incompatible Benchmarks**
- **Black-Box Behavior**
- **Defective Hardware**
- **Hostile Environmental**

Conclusion

- State-of-the-art mitigation methods have vital drawbacks:
 - Accuracy-vs-Safety trade-off, High resource/time demand, Requirement of expert knowledge, etc.
- A demand to move away from traditional broad-spectrum standardization
- Utilizing safety concern lists to create a safety case argument for the application