Technical Report: AI Safety Models Proof of Concept (POC)

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Project: AI Safety Models for Conversational Platforms

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1. High-Level Design Decisions

The objective of this POC is to implement a suite of AI safety models that enhance user safety in a conversational platform. The key requirements include:

- 1. **Abuse Language Detection** Real-time detection of harmful, threatening, or inappropriate content.
- 2. **Escalation Pattern Recognition** Detecting conversations becoming emotionally dangerous.
- 3. Crisis Intervention Recognition of severe emotional distress or self-harm indicators.
- 4. **Content Filtering** Age-appropriate filtering for guardian-supervised accounts.

Design Approach:

- Use **Python** as the primary language.
- Utilize **Hugging Face Transformers (BERT)** for advanced abuse detection.
- Implement a baseline Logistic Regression model with TF-IDF features for comparative predictions.
- Integrate both models into a **Streamlit-based web simulator** for real-time or near-real-time analysis.
- Include session-based escalation detection, crisis keyword alerts, and age-appropriate content filtering.

2. Project Structure

```
AI safety Jasmine/
— data/
                     # Raw CSV datasets
    - raw/
     - processed/
                       # Train/test split CSVs
  - models/
  baseline model.pkl # Logistic Regression + TF-IDF
  · src/
    - datapreprocessing.py # Data cleaning, encoding, splitting
    train model.py
                        # Baseline model training
                    # Streamlit app integrating all models
  app.py
 - requirements.txt
                        # Python dependencies
 - README.md
                         # Project overview and instructions
 - docs/
 architecture.png # Architecture diagram (placeholder)
```

3. Data Sources and Preprocessing

• **Dataset:** Publicly available anonymized conversational text with labels: safe and unsafe.

• Preprocessing Steps:

- 1. Remove missing/NaN entries.
- 2. Clean text: lowercase, remove punctuation and unnecessary characters.
- 3. Encode labels into numeric values (0=safe, 1=unsafe).
- 4. Split into training (80%) and testing (20%) sets.

Tools: pandas, scikit-learn

4. Model Architectures and Training Details

4.1 Baseline Model

- **Pipeline:** TF-IDF vectorizer → Logistic Regression classifier
- **TF-IDF Parameters:** max_features=5000, ngram_range=(1,2)

- Training: Standard CPU-based training
- Evaluation Metrics: Precision, Recall, F1-Score

4.2 BERT Model

- Architecture: Pretrained bert-base-uncased transformer for text classification
- Inference: Returns probability scores for safe and unsafe labels
- Integration: Used alongside baseline model to improve decision-making

4.3 Ensemble Decision

- If both baseline and BERT models agree, use that label
- If disagreement, baseline model acts as tie-breaker (for simplicity in POC)

4.4 Escalation Detection

- Monitors the last 3 consecutive unsafe messages
- Triggers a warning if all last three messages are unsafe

4.5 Crisis Intervention

- Detects presence of keywords like "kill myself", "suicide", "hopeless"
- Triggers emergency alert in the simulator

4.6 Age-Appropriate Filtering

- Child, teen, adult categories
- Each has a predefined list of restricted keywords
- Alerts if user input violates age restrictions

5. Evaluation Results

Baseline Model Evaluation:

Class	Precisio n	Recall	F1-Score	Support
Safe	0.00	0.00	0.00	1
Unsafe	0.50	1.00	0.67	1
Accuracy			0.50	2
Macro Avg	0.25	0.50	0.33	2
Weighted Avg	0.25	0.50	0.33	2

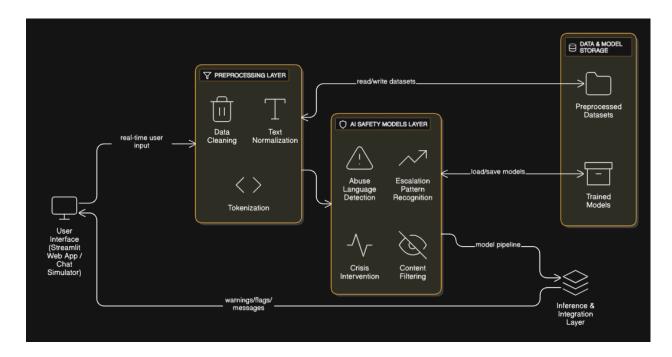
Notes:

- Small test set results in unstable metrics.
- UndefinedMetricWarning handled with zero_division=0.
- Focus of POC is **functionality demonstration**, not production accuracy.
- Larger datasets required for improved performance.

6. Leadership and Project Management Considerations

- Modular Design: Easy to extend models or integrate with APIs.
- **Bias Mitigation:** Preprocessing and thresholding reduce extreme misclassifications.
- **Team Guidance:** Emphasis on code modularity, proper testing, and dataset expansion.
- Ethical Considerations: Careful handling of crisis messages and age-restricted content.

7. Architecture Diagram



Description:

- User sends a message via Streamlit interface
- Baseline + BERT models analyze text
- Ensemble decision determines safety label
- Escalation detection monitors recent messages
- Crisis alerts trigger if high-risk keywords are detected
- Age filtering enforces restrictions based on user group

8. Future Improvements

- 1. Expand dataset to include multi-language, slang, and edge cases
- 2. Fine-tune BERT on domain-specific abusive datasets

- 3. Implement advanced ensemble methods (weighted average, stacking)
- 4. Add real-time API integration for chat platforms
- 5. Enhance user interface with historical conversation tracking