

# Methodological Expo I — ME–Data Track

(20–25 minutes total)

## Key themes:

- **Dataset description**
  - Name, source, size, time range etc.
- **Why this data?**
  - What RQs motivated the use of this data?
    - This should be problem-centered and not data-centered here!
  - What could have been alternatives, and why not those?
  - What makes this data to be unique?
- **Data collection:**
  - Scraping/API/Existing public data released, etc.
  - Permissions/rate limits/access requests
  - Any restrictions?
  - Metadata collected
  - Program/code?
  - Storage formatting
  - Manual steps involved?
  - Show code snippets, flow diagram, collection pipeline here.
- **Challenges faced:**
  - API issues, missing fields, large file handling, computational limitations etc.
  - Bots/spams/noise etc.
- **How did you address the challenges?**
  - Any data cleaning?
  - Any scripts to handle errors?
- **Data cleaning and preprocessing?**
  - Give snapshots of raw data and final data
  - What all was removed and why?
    - E.g., removal of duplicates, missing values handling, removal of non-English content, normalizing timestamps, etc.
- **Reflection:**
  - Ethics of the data collection
  - Potential bias and risks
    - E.g., self-selection, platform demographic skew, etc.
  - Potential misuse of data/ bad actors
  - Reproducibility
    - Can this data be shared broadly?
    - Can the collection and processing scripts be shared publicly?
  - Other interesting comments about the data?

# Methodological Expo I — ME-MethodsTrack

(20–25 minutes total)

## Key themes:

- **Method description**
  - Name of the method
  - Type of method (statistical / ML / NLP / network / causal / qualitative-computational hybrid, etc.)
  - What problem does this method solve?
  - What type of data does it operate on?
- **Why this method?**
  - What research question motivated the use of this method?
  - Why is this method appropriate for the problem?
  - What could have been alternative methods?
  - Why were those alternatives not chosen?
  - What makes this method particularly useful or powerful here?
    - This should be problem-centered, not method-for-the-sake-of-method.
- **How the method works**
  - Explain clearly and step-by-step:
    - Conceptual intuition behind the method
    - Mathematical or algorithmic logic
    - Inputs required
    - Parameters or hyperparameters
    - Output produced
    - How results are interpreted
  - Show:
    - Workflow diagram/Pseudocode or simplified code snippet/Model architecture (if applicable)/Formula (if statistical method)
    - The goal is clarity, not technical jargon only.
  - **Implementation**
    - What tools/libraries were used?
    - What programming language?
    - Any preprocessing required before applying the method?
    - How long did it take to run?
    - Computational requirements?
    - Any parameter tuning?
    - Show: Code snippet/Pipeline diagram/Example outputs
  - **Challenges faced**
    - Convergence issues, Model instability, Overfitting, Class imbalance, Computational limits, Parameter sensitivity

- Interpretability, Choosing thresholds
- Performance evaluation
- Ground truth limitations
- How did you address the challenges?
  - Cross-validation? Regularization? Alternative specifications? Robustness checks?
- **Evaluation:**
  - How did you assess performance or validity?
  - What metrics were used? (Accuracy, F1, RMSE, AUC, etc.)
  - Any baselines used for comparison?
  - Any robustness checks?
  - Any sensitivity analysis?
- **Reflection:**
  - Limitations and assumptions
  - Bias and fairness concerns
  - Ethical concerns:
    - Risk of misuse
    - Overclaiming causality or predictive power
  - **Reproducibility:**
    - Can code be shared?
    - Are results deterministic?
    - Any randomness or seed sensitivity?
  - **Other interesting comments about the method?**
    - When should someone NOT use this method?
    - How would this scale to larger datasets?
    - Would it generalize to another platform/domain?