



Hackathon Judging Criteria

Your project will be evaluated in **2 main areas**:

- 🎯 **Model Performance (60%)**
 - ⚙️ **Practical Considerations (40%)**
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1 Primary Metrics (60%)

These measures how well your model predicts who will get sick.



F2-Score (Most Important)

- Focuses more on **recall than precision**
- Catching sick people is more important than avoiding false alarms
- Missing a sick person = worse than flagging a healthy one



Your model should prioritize identifying at-risk individuals.



PR-AUC (Precision-Recall AUC)

- Best for **imbalanced datasets**
- Most people are healthy → very few become sick
- Measures performance when positive cases are rare



Your model must work well even when disease cases are limited.



ROC-AUC

- Industry standard metric
 - Helps compare your model with published research
 - Measures overall classification ability
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2 Additional Criteria (40%)

These focus on real-world practicality and ethics.

No Data Leakage

You cannot use features that already reveal the disease.

Example:

- If someone is taking diabetes medication → they already have diabetes
- Using that feature = cheating

 Judges will audit your feature set.

Real-World Usability

Your model must handle:

- Noisy inputs
- Missing values
- Inconsistent self-reported data

Remember:

You're not working with perfect hospital records — only self-reported data.



Cost Efficiency

- Simple > Complex
- Lightweight models > Heavy API-based systems
- Fast runtime is better

Avoid overengineering.



Open Source Only

- All tools must be open-source
 - Code must be reproducible
 - No proprietary black-box APIs
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Explainability

You must explain:

- Why was this person flagged?
- Which features contributed most?

Both:

- Users
- Doctors

should understand the reasoning.



Fairness

Your model must NOT discriminate based on:

- Age

- Gender
- Ethnicity

Judges will check for bias.

Bonus Points

Extra credit for:

- ✨ Creative feature engineering
 - ✨ Discovering non-obvious correlations
 - ✨ Clean, production-ready code
 - ✨ Clear documentation
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What This Means Strategically

To win:

1. Maximize **F2-score**
2. Avoid leakage
3. Keep model simple & explainable
4. Handle missing/noisy data well
5. Show fairness checks
6. Keep everything open-source