Best Practices for Competition Success

STEP 1: UNDERSTAND THE PROBLEM STATEMENT

- Clarify Objectives & Business Impact: Determine whether it's a classification or regression task, the domain context, and what "success" looks like.
- Target Variable: Understand how the target is defined and distributed.
- Constraints & Permissions: External data limits, privacy, or regulatory issues.
- Action Points:
 - Annotate key requirements and assumptions.
 - Research the domain for feature ideas.
 - Restate the problem in your own words to ensure clarity.

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STEP 2: STUDY THE EVALUATION METRIC

- Metric Matters: Each competition or project has a specific metric (e.g., F1-score, AUC-ROC, RMSE), which guides how you tune and compare models.
- Classification vs. Regression Metrics:
 - ullet Accuracy, F1-score, AUC-ROC, and Log Loss for classification. $oldsymbol{\lor}$
 - RMSE, MAE, R² for regression.
- Optimization Direction: Decide if you must maximize or minimize the metric; some metrics (e.g., AUC-ROC) may need a proxy loss for training.
- Action Points:
 - Simulate prediction changes to see impact on the metric.
 - If allowed, implement custom losses aligned to the competition metric.

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STEP 3: PERFORM AN INITIAL DATA ANALYSIS (EDA)

- Data Profiling: Summarize shape, basic statistics, feature and target distribution
- Quality Checks: Identify missing values, outliers, or suspicious patterns; confirm no hidden data leakage.
- Train-Test Shift: Check if train and test set differ significantly in their feature distributions.
- Action Points:
 - Use tools like Pandas Profiling or Sweetviz for automated EDA.
 - Document anomalies or domain-specific oddities.
 - Examine correlations, scatter plots, and histograms.

STEP 4: DEVELOP A BASELINE MODEL

- Why a Baseline? A simple model (e.g., Logistic Regression, Decision Tree) gives you a reference performance.
- Pipeline Check: Make sure that data loading, preprocessing, and validation splits are correct.
- Diagnostics: Analyze misclassifications/ residuals to spot improvement areas.
- Action Points:
 - Start with a naive or default-parameter model.
 - Use cross-validation to assess reliability.
 - Record baseline metrics to quantify future gains.

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STEP 5: SET UP A ROBUST VALIDATION STRATEGY

- Importance of Validation: Avoid overfitting or a single random split.
- Common Methods:
 - K-Fold (balanced data).
 - Stratified K-Fold (imbalanced classes).
 - Time-Based or Group Splits (time series, repeated measures).
- Action Points:
 - Match validation approach to the final test scenario.
 - Keep a holdout set for final checks.
 - Use consistent folds across experiments to compare fairly.

STEP 6: FEATURE ENGINEERING AND DATA PREPROCESSING

- Missing Data: Mean/median imputation, advanced methods, or flags.
- Encoding Categorical Variables: One-hot/ target/ frequency enc. (leakage?!).
- Scaling & Transformations: Normalize or log-transform skewed features, especially for sensitive models (SVMs, neural nets).
- **New Features**: Time-based, domain knowledge, interactions (e.g. ratios).
- Action Points:
 - Add features in small batches and track validation changes.
 - Use scikit-learn Pipelines to avoid leakage.
 - Consider dimensionality reduction if feature space is large.

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STEP 7: SELECT AND TUNE MODELS STRATEGICALLY

Model Choices:

- Tree-based ensembles (XGBoost, LightGBM, CatBoost) for tabular data.
- Neural networks for large-scale or unstructured data.
- Simple linear/logistic models can still compete with good feature engineering.

• Hyperparameter Tuning:

- Grid Search, Random Search, Bayesian Optimization.
- Early stopping and regularization (L1, L2) to avoid overfitting.

Action Points:

- Track experiments (MLflow, Weights & Biases, or spreadsheets).
- Start with defaults and refine gradually.
- Use IML tools (SHAP, permutation importance) for feature insights.

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STEP 8: TRACK EXPERIMENTS AND ITERATE

- Experiment Logging: Keep detailed records of data versions, hyperparams, code revisions.
- Error Analysis: Study misclassifications and residuals to guide feature tweaks.
- Iterative Improvements:
 - Refine features, adjust validation, revisit transformations.
 - Maintain a "changelog" for each iteration.
- Action Points:
 - Automate repeated tasks (submissions, data prep).
 - Prune ineffective ideas swiftly.
 - Ensure reproducibility (version control, environment snapshots).

STEP 9: UNDERSTAND THE LEADERBOARD AND AVOID OVERFITTING

- Public vs. Private Leaderboard: Public boards use only part of the test set—overfitting leads to dramatic rank changes later.
- Submission Strategy:
 - Rely mainly on internal validation.
 - Keep a private holdout to confirm final performance.
- Leaderboard Shakeups: Watch for big changes when private scores are revealed; it often indicates overfitting to the public set.
- Action Points:
 - Limit minor tweaks solely to improve a public rank.
 - If a jump seems suspicious, re-check data handling for leakage.

STEP 10: COLLABORATE AND LEARN FROM OTHERS

 Study Top Solutions: Many winning approaches are documented in blog posts or Kaggle discussions.

Teamwork:

- Collaboration brings diverse perspectives and model ensembling.
- Share code and features for synergy.

Community & Networking:

- Discussion forums reveal dataset quirks, best practices.
- Local meetups or online communities keep you updated.

Action Points:

- Document and share your findings.
- Adapt others' ideas carefully—validate them on your own splits.
- Engage with peers for new tools and techniques.

FINAL ADVICE

- Start Simple, Then Iterate: Don't jump into complex models, get a solid baseline.
- **Prioritize Validation**: A robust validation strategy prevents nasty surprises.
- **Log Everything**: Document transformations, parameters, and model versions.
- Aim for Generalizable Solutions: Don't just chase leaderboard scores—focus on reproducibility and reliability.