ARC PRIZE 2025 RESEARCH-DRIVEN SOLVER DEVELOPMENT PROMPT

OVERVIEW AND MISSION

You are tasked with developing an AI systems to efficiently learn new skills and solve openended problems, rather than depend exclusively on systems trained with extensive datasets for the ARC Prize 2025 (Abstraction and Reasoning Corpus) competition on Kaggle. This is a research-driven development process where you will:

- 1. **Comprehensively review** all provided research documents and literature
- 2. **Synthesize** state-of-the-art approaches from the research
- 3. **Design** a novel or hybrid solution architecture based on your analysis
- 4. **Implement** the solution through iterative refinement
- 5. **Deliver** a production-ready Kaggle notebook (.py file)

Critical Principle: Your approach must be **derived from research**, not predetermined. You will analyze the literature, identify promising directions, and make informed architectural decisions based on empirical evidence and theoretical foundations from the research community.

EXECUTION MODE: This is a single-pass, complete workflow. Execute all steps sequentially without pausing. For each step, produce the specified deliverables in full before proceeding to the next step.

KAGGLE ENVIRONMENT:

- Your .py file is the solution code (what you submit to Kaggle)
- Input data path: `/kaggle/input/arc-prize-2025/arc-agi test challenges.json`
- Your code must write: `submission.json` in current directory
- Do NOT reference training/evaluation/solution files in the final .py code
- The .py code must be self-contained and work with only the test challenges ison file

STEP 1: COMPREHENSIVE RESEARCH ANALYSIS AND PROBLEM UNDERSTANDING

Prompt:

Conduct a thorough analysis of the ARC Prize 2025 challenge by synthesizing all available resources in this directory.e.g everything in /home/legend/Documents/AGI/Kaggle, and sub folders like /home/legend/Documents/AGI/Kaggle/literature:

- 1. **Primary Problem Documents Analysis:**
 - Read ALL provided files about the ARC Prize 2025
 - Extract the exact problem specification:
 - * Task structure (JSON format, train/test pairs)
 - * Grid constraints (dimensions, color values)
 - * Submission format requirements (pass@2, attempt_1/attempt_2)
 - * Scoring methodology (exact matching)
 - Identify competition constraints:
 - * Computational limits (12-hour runtime, no internet access)
 - * Hardware specifications (GPU availability, memory)
 - * Kaggle-specific requirements (file paths, deterministic execution)
- 2. **Existing Research Literature Review:**
- Analyze the solutions so far and research documents provided (e.g., "/home/legend/Documents/AGI/Kaggle/literature/papers, /home/legend/Documents/AGI/Kaggle/arc_prize_2025_submission and sun folders, /home/legend/Documents/AGI/Kaggle/literature, any paper, code, approach, etc)
 - Extract key methodologies mentioned: e.g
 - * Domain-Specific Language (DSL) approaches
 - * Program synthesis techniques
 - * LLM-augmented methods
 - * Object-centric neuro-symbolic approaches
 - Identify what has worked historically:
 - * Performance benchmarks from literature and online searches
 - * Winning solutions from previous competitions
 - * Theoretical foundations cited
- 3. **Online Research Supplementation:**

Use web_search to gather:

- Recent papers on ARC solving (2023-2025)
- Kaggle discussion forums for ARC Prize 2025
- Current leaderboard standings and publicly shared approaches

- GitHub repositories of successful ARC solvers
- Blog posts and technical write-ups from competition participants

B. Technical Problem Decomposition

Based on your research, answer:

- 1. **What makes ARC tasks difficult?**
 - Core cognitive requirements
 - Why traditional ML/LLMs fail
 - Specific reasoning challenges
- 2. **What are the latest best proven solution paradigms?**
 - Categorize approaches from literature (e.g symbolic, neural, hybrid etc)
 - Identify success rates and limitations of each
 - Note computational trade-offs
- 3. **What are the key design decisions?**
 - DSL expressiveness vs. search space size
 - Search strategy (e.g breadth-first, beam search, Monte Carlo etc)
 - Verification mechanisms
 - Heuristic design
 - Two-attempt strategy formulations

C. Comparative Analysis

Create a detailed comparison table:

- List 5-7 major approaches from literature
- For each: core principle, performance, computational cost, implementation complexity
- Identify gaps or opportunities for hybrid approaches

DELIVERABLE:

Research synthesis document (800-1000 words) including:

- Problem specification summary (200 words)
- Key findings from literature (300 words)
- Comparative analysis table (3-5 approaches)
- Preliminary architectural recommendation (200 words)

STEP 2: ARCHITECTURAL DESIGN AND APPROACH SELECTION

Prompt:

Based on your research synthesis, design a solver architecture:

A. Architecture Selection Rationale example: Use better one if you find a better one in your research

1. **Choose Your Core Approach:**

Justify your selection using research evidence:

- Will you use pure DSL-based program synthesis?
- Hybrid neuro-symbolic?
- LLM-augmented synthesis?
- Novel combination?

For your choice, provide:

- Supporting evidence from literature (cite specific papers/results)
- Why this approach suits ARC Prize 2025 constraints
- Expected performance range based on benchmarks
- Known limitations and mitigation strategies
- 2. **Component Architecture Design:**

Define the major system components:

- **Perception/Analysis Layer**: How will you analyze training pairs?
- **Reasoning/Synthesis Layer**: How will you generate candidate solutions?

- **Verification Layer**: How will you validate candidates?
- **Execution Layer**: How will you apply solutions to test inputs?

For each component:

- Technical approach (specific algorithms/methods)
- Research citations supporting the approach
- Expected computational complexity
- Failure modes and handling strategies

3. **Two-Attempt Strategy Design:**

Design how you'll generate diverse attempts:

- What defines "diversity" in your approach?
- How will attempt_1 and attempt_2 differ?
- Research evidence for ensemble/multi-strategy benefits

B. Implementation Specification

Create detailed specifications WITHOUT writing code:

- 1. **DSL Design (if applicable):**
 - List of primitive operations (justified by task analysis)
 - Operation categories (geometric, color, structural, etc.)
 - Parameter spaces for each operation
 - Composability rules

2. **Search Strategy:**

- Algorithm choice (beam search, A*, MCTS, etc.)
- Search space pruning heuristics
- Stopping criteria
- Computational budgets per task

- 3. **Heuristic Design:**
 - Feature extraction from training pairs
 - How features guide search
 - Priority ordering mechanisms
 - Adaptation strategies
- 4. **Verification Mechanism:**
 - Exact matching on training pairs
 - Handling of ambiguous cases
 - Confidence scoring (if applicable)

C. Theoretical Analysis

- 1. **Completeness Analysis:**
- What percentage of ARC tasks is your approach theoretically capable of solving?
 - What categories of tasks will it struggle with?
- 2. **Computational Feasibility:**
 - Estimated time per task
 - Scalability to 100+ tasks
 - Memory requirements
- 3. **Expected Performance:**
 - Predicted accuracy range based on literature benchmarks
 - Comparison to current leaderboard (from online research)
 - Conservative vs. optimistic estimates

DELIVERABLE:

Architecture design document (600-800 words):

- Approach selection rationale with citations (300 words)
- Component specifications (200 words)

- Theoretical performance analysis (150 words)

STEP 3: ITERATION 1 - INITIAL IMPLEMENTATION

Prompt:

Implement your first version of the solver based on your architectural design:

A. Implementation

Write a complete, production-ready Python file that:

- 1. Implements ALL components from your architecture
- 2. Loads data from Kaggle paths: `/kaggle/input/arc-prize-2025/arc-agi_test_challenges.json`

Guve kaggles notebook starter as provided by Kaggle {# This Python 3 environment comes with many helpful analytics libraries installed

It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-python

For example, here's several helpful packages to load

import numpy as np # linear algebra

import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)

Input data files are available in the read-only "../input/" directory

For example, running this (by clicking run or pressing Shift+Enter) will
list all files under the input directory

import os

for dirname, _, filenames in os.walk('/kaggle/input'):

for filename in filenames:

print(os.path.join(dirname, filename))

You can write up to 20GB to the current directory (/kaggle/working/) that gets preserved as output when you create a version using "Save & Run All"

- # You can also write temporary files to /kaggle/temp/, but they won't be saved outside of the current session}
- 3. Generates valid `submission.json` with correct format
- 4. Handles all edge cases gracefully
- 5. Runs deterministically (set seeds)
- 6. Includes clear code structure and comments
- **Implementation Guidelines:**
- Follow your architecture specification exactly
- Implement complete functionality (no stubs or TODOs)
- Use only libraries available on Kaggle (numpy, scipy, no internet-dependent packages)
- Include error handling and fallback mechanisms
- Ensure computational efficiency where possible

B. Theoretical Correctness Analysis

Conduct focused self-review (300-400 words total):

- Logic verification: Walk through 2-3 example tasks mentally
- Compliance check: Verify all Kaggle requirements
- Critical assessment: 3 key strengths, 3 key weaknesses, performance estimate

DELIVERABLE:

- `arc_solver_v1.py`
- Self-assessment (300-400 words)
- 2. **Mental Task Walkthrough:**

Select 3 example task types (from your understanding of ARC):

- Simple pattern (e.g., color remapping)
- Geometric transformation (e.g., rotation + scaling)
- Complex compositional (e.g., multi-rule interaction)

For each, trace through your code mentally:

- Would it extract correct features?
- Would it generate appropriate candidates?
- Would verification work correctly?
- Would it produce valid output?
- 3. **Research Benchmark Comparison:**
 - Compare your approach to methods in literature
 - Identify where your implementation might underperform
 - Identify potential advantages

C. Compliance and Feasibility Review

- 1. **Competition Requirements:**
 - [] Correct input file path
 - [] Correct submission.json format
 - [] Two attempts per test input
 - [] All tasks covered (no missing task_ids)
 - [] Grids are valid (0-9 values, reasonable dimensions)
 - [] No internet access required
 - [] Deterministic execution
- 2. **Computational Complexity Analysis:**
 - Estimate runtime per task (worst case, average case)
 - Calculate total expected runtime for 100 tasks
 - Identify computational bottlenecks
 - Assess if it fits in 12-hour limit
- 3. **Code Quality:**
 - Readability and maintainability
 - Proper error handling

- Modularity and extensibility

D. Critical Self-Assessment

Write an honest critique:

- **Strengths**: What does this implementation do well?
- **Weaknesses**: Where is it likely to fail?
- **Theoretical Performance**: Estimate accuracy (e.g., "5-15%")
- **Comparison to SOTA**: How far from competitive solutions?
- **Key Improvements Needed**: Prioritized list

DELIVERABLE:

- `arc_solver_v1.py` (complete, runnable file)
- Self-assessment document (800-1200 words)
- Identified improvements for next iteration

STEP 4: ITERATION 2 - REFINEMENT AND ENHANCEMENT

Prompt:

Refine your solver based on V1 analysis and additional research:

A. Gap Analysis and Research

- 1. **Review V1 Weaknesses:**
 - Identify the top 3-5 critical issues from V1 assessment
 - For each issue, conduct targeted research:
- * Use web_search for specific solutions (e.g., "ARC task color symmetry detection")
 - * Look for papers addressing these specific challenges
 - * Search Kaggle discussions for practical solutions

- 2. **Enhancement Opportunities:**
 - Research advanced techniques not in V1:
 - * More sophisticated search strategies
 - * Better heuristics
 - * Additional DSL operations
 - * Improved verification methods
 - Find specific algorithmic improvements from recent work

B. Implementation V2

Create an improved version:

- 1. **Address Critical Issues:**
 - Fix logical errors from V1
 - Improve components that were weak
 - Add missing functionality
- 2. **Add Enhancements:**
 - Implement research-backed improvements
 - Expand DSL (if applicable) with new operations
 - Improve search efficiency
 - Better feature extraction
- 3. **Maintain Strengths:**
 - Keep working components from V1
 - Preserve successful design decisions
 - Ensure compliance still met

C. Advanced Evaluation

1. **Cross-Validation Against Research:**

- Compare V2 approach to 3-5 recent papers
- Identify which techniques from literature you've incorporated
- Note which proven techniques you're still missing
- 2. **Theoretical Task Coverage:**
 - Categorize ARC task types (from problem understanding)
 - For each category, assess: Can V2 solve it?
 - Estimate coverage: "V2 should handle X% of Y-type tasks"
- 3. **Computational Profile:**
 - Re-analyze runtime complexity
 - Compare to V1: Is it faster/slower?
 - Is it still within 12-hour budget?
- 4. **Online Benchmarking:**
 - Use web_search to find:
 - * Current Kaggle leaderboard scores
 - * Discussion of what scores are competitive
 - * Public solution approaches and their performance
 - Compare V2's expected performance to these benchmarks
- ### D. Critical Assessment V2
- 1. **Progress Evaluation:**
 - Quantify improvements over V1
 - Justify why changes should help
 - Estimate new performance range
- 2. **Remaining Gaps:**
 - What still doesn't work?
 - What tasks will V2 still fail on?

3. **Path to V3:**

- Identify THE most impactful improvement for V3
- Research backing for this improvement
- Implementation strategy

DELIVERABLE:

- `arc_solver_v2.py`
- V1→V2 improvement summary (400-500 words)
- Performance estimate with justifications

STEP 5: ITERATION 3 - OPTIMIZATION AND ROBUSTNESS

Prompt:

Create the most robust and competitive version before finalization:

A. State-of-the-Art Alignment Research

- 1. **Deep Dive into Top Approaches:**
 - Use web_search to find:
 - * Recent competition winners' approaches (2024-2025)
 - * Highest-performing published methods
 - * Novel techniques from latest papers
 - Extract specific implementation details
 - Identify "secret sauce" components
- 2. **Competitive Gap Analysis:**
- Current leaderboard scores (from online research e.g https://www.kaggle.com/competitions/arc-prize-2025/leaderboard)
 - Your V2 estimated performance

- Gap size and components causing the gap
- Feasibility of closing gap with V3

B. Implementation V3

Final pre-production version:

- 1. **Incorporate SOTA Techniques:**
 - Add most promising technique from research
 - Implement sophisticated optimizations
 - Enhance robustness and edge case handling
- 2. **Optimization Pass:**
 - Profile computational bottlenecks (theoretically)
 - Implement caching where beneficial
 - Optimize critical paths
 - Balance quality vs. speed
- 3. **Robustness Hardening:**
 - Comprehensive error handling
 - Graceful degradation for hard tasks
 - Ensure NO crashes regardless of input
 - Validate all outputs before submission

C. Comprehensive Evaluation

1. **Multi-Dimensional Analysis:**

Correctness:

- Mental walkthrough of 5+ diverse task types
- Verify logic for each component

- Check for subtle bugs or edge cases
- **Compliance:**
- Re-verify all Kaggle requirements
- Check submission format rigorously
- Confirm deterministic behavior
- **Performance:**
- Estimate accuracy on:
 - * Simple tasks (pattern matching)
 - * Medium tasks (2-3 operation compositions)
 - * Complex tasks (4+ operations, interactions)
- Overall accuracy estimate
- **Efficiency:**
- Task time distribution (fast/slow tasks)
- Total expected runtime
- Memory usage assessment
- 2. **Research-Based Validation:**
 - Compare V3 to each major approach from Step 1 literature review
 - For each: What techniques did you adopt? What did you skip? Why?
 - Identify your solver's unique contributions or combinations
- 3. **Online Competitive Analysis:**
 - Latest Kaggle leaderboard standings
 - Discussion forum insights
 - Comparison: Where does V3 likely rank?
 - What would push it higher?
- ### D. Failure Mode Analysis

- 1. **Identify Task Categories V3 Will Fail:**
 - Specific examples from ARC task taxonomy
 - Why will it fail? (insufficient DSL, search timeout, etc.)
 - Could these be addressed? At what cost?

2. **Known Limitations:**

- Computational constraints causing shortcuts
- Theoretical limitations of approach
- Trade-offs made and their implications

3. **Risk Assessment:**

- Probability of submission errors
- Probability of timeout
- Expected score variance (best/worst case)

E. V3 Critical Assessment

Final honest evaluation:

- **Architecture Quality**: Is the design sound?
- **Implementation Quality**: Is the code production-ready?
- **Expected Performance**: Realistic accuracy estimate with reasoning
- **Competitive Standing**: Where in leaderboard range?
- **Confidence Level**: High/Medium/Low and why
- **Key Strengths**: Top 3 advantages
- **Key Weaknesses**: Top 3 limitations

DELIVERABLE:

- `arc_solver_v2.py`
- V1→V2 improvement summary (400-500 words)
- Performance estimate with justification

STEP 6: FINAL VERSION - PRODUCTION IMPLEMENTATION

Prompt:

Create the final, production-ready solver for Kaggle submission:

A. Final Implementation Decision

Based on V3 assessment:

- 1. **If V3 is strong**: Polish and finalize it
- 2. **If critical issue found**: Make targeted fix, creating V3.1
- 3. **If major rework needed**: Explain what and why, then implement

B. Production-Ready Code

Create `arc_solver_final.py`:

- 1. **Code Quality:**
 - Clean, well-commented code
 - Proper structure and organization
 - Professional naming conventions
 - Comprehensive docstrings
- 2. **Robustness:**
 - Handles ALL edge cases
 - Never crashes or hangs
 - Always produces valid submission
 - Graceful degradation for hard tasks
- 3. **Kaggle Optimization:**
 - Efficient imports (only necessary libraries)

- Proper file paths for Kaggle environment
- Deterministic with fixed random seeds
- Appropriate logging/progress indicators

4. **Submission Guarantee:**

- Always generates valid submission.json
- Correct format for every task
- Two attempts per test input
- No missing or malformed entries

C. Final Validation

1. **Mental Execution Test:**

- Trace through the complete pipeline
- Verify data flow from input to submission
- Check all branches and conditions
- Confirm error handling paths work

2. **Requirements Checklist:**

- [] Correct Kaggle input path
- [] Valid submission.json output
- [] Pass@2 format (two attempts)
- [] No internet dependencies
- [] Deterministic execution
- [] Runs within time limit (estimated)
- [] Handles all 100+ tasks
- [] Values in [0-9] range
- [] Grid dimensions valid (1x1 to 30x30)

Final Complexity Analysis:

- Best case runtime

- Average case runtime
- Worst case runtime
- Confidence in 12-hour completion

D. Documentation Package

Create concise documentation:

- 1. **Approach Evolution** (600 words):
 - Research foundation and key decisions (250 words)
 - Version progression V1→V2→V3→Final (250 words)
 - Implementation highlights (100 words)
- 2. **Final Approach Explanation** (600 words):
 - Architecture and core components (250 words)
 - Algorithmic strategy and two-attempt design (200 words)
 - Expected performance and limitations (150 words)

DELIVERABLE:

- `arc_solver_final.py`
- `approach_documentation.md` (both sections combined, ~1200 words total)

STEP 7: FINAL RESEARCH REFLECTION AND SUBMISSION PACKAGE

Prompt:

STEP 7: FINAL PACKAGE

Organize all deliverables with brief summary:

- Final performance prediction (confidence intervals)
- List of all 4 Python files produced
- 200-word reflection on research-to-implementation journey

DELIVERABLE:

- Summary document (3000 words)