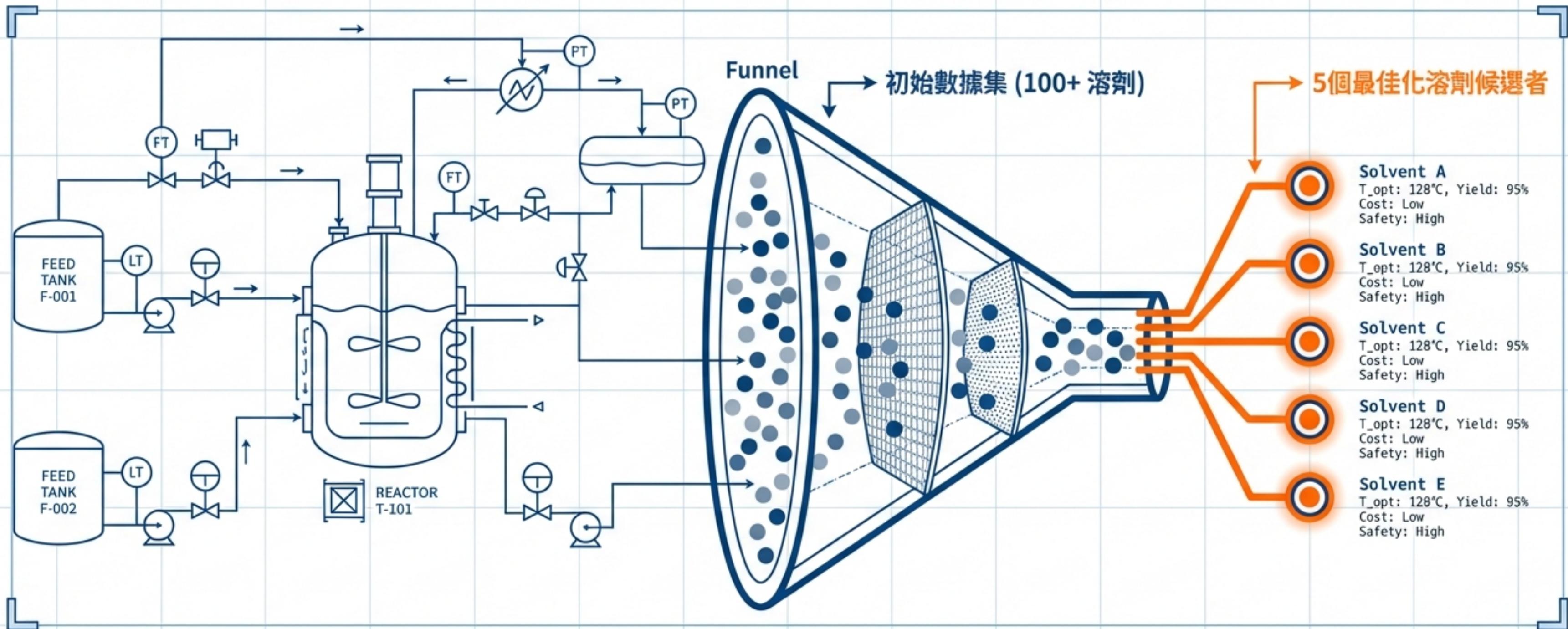


AI 在化工上的應用：溶劑篩選案例研究

Unit 09 - 從 100 到 5：非監督式學習的綜合應用



溶劑篩選的挑戰：多目標衝突與高昂成本

多重物理限制 (Conflicts)



- 高溶解度 (High Solubility) vs 低毒性 (Low Toxicity)
- 低沸點 (Low Boiling Point) vs 低揮發損耗 (Low Loss)
- 高性能 (High Performance) vs 低成本 (Low Cost)

方法比較 (Paradigm Shift)



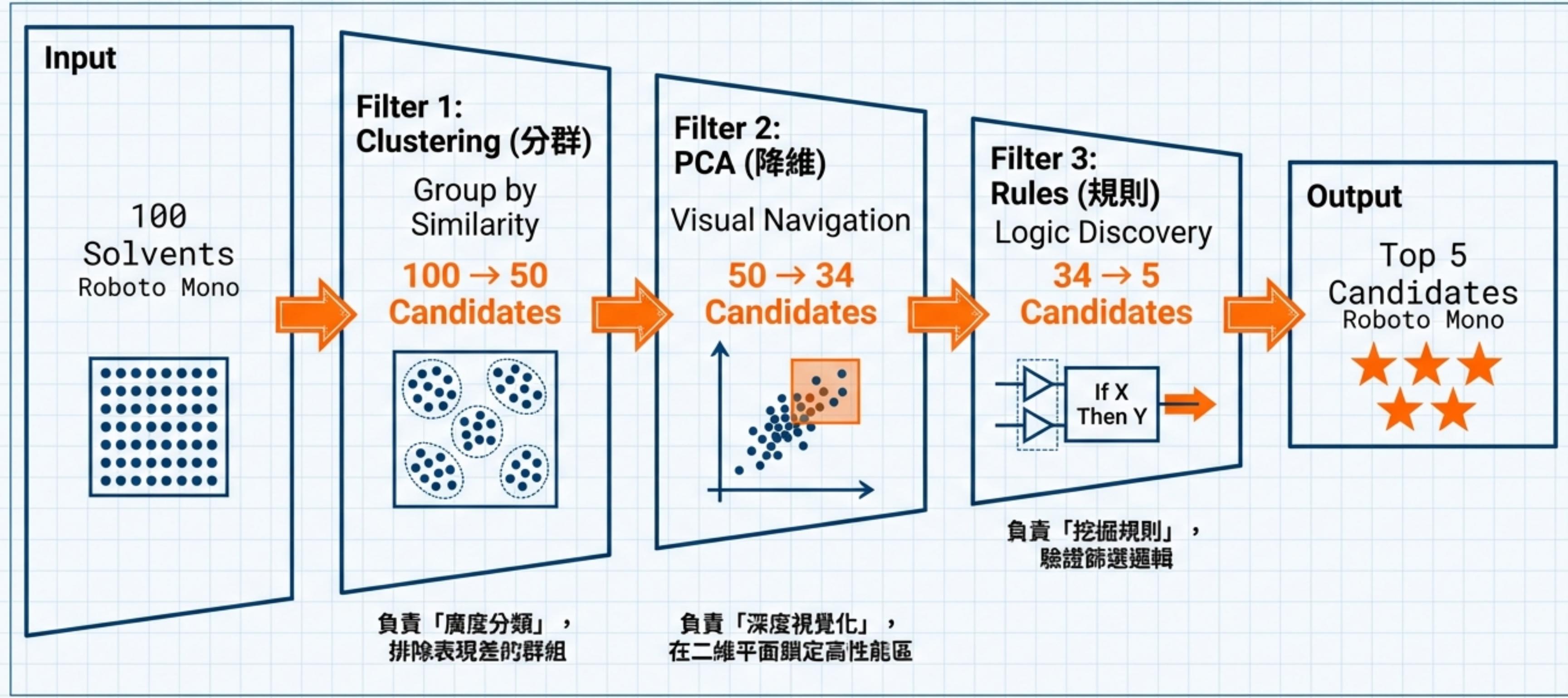
傳統方法 (Traditional)
試誤法 (Trial & Error) - 高成本、
依賴直覺、易受主觀影響。



AI 方法 (AI Approach)
數據驅動、多維度評估、**快速篩選 (Fast Screening)**。

核心目標：如何在不進行 100 次實驗**的情況下，找出那 5 個最佳候選？**

AI 篩選漏斗：三種非監督式學習的協同效應

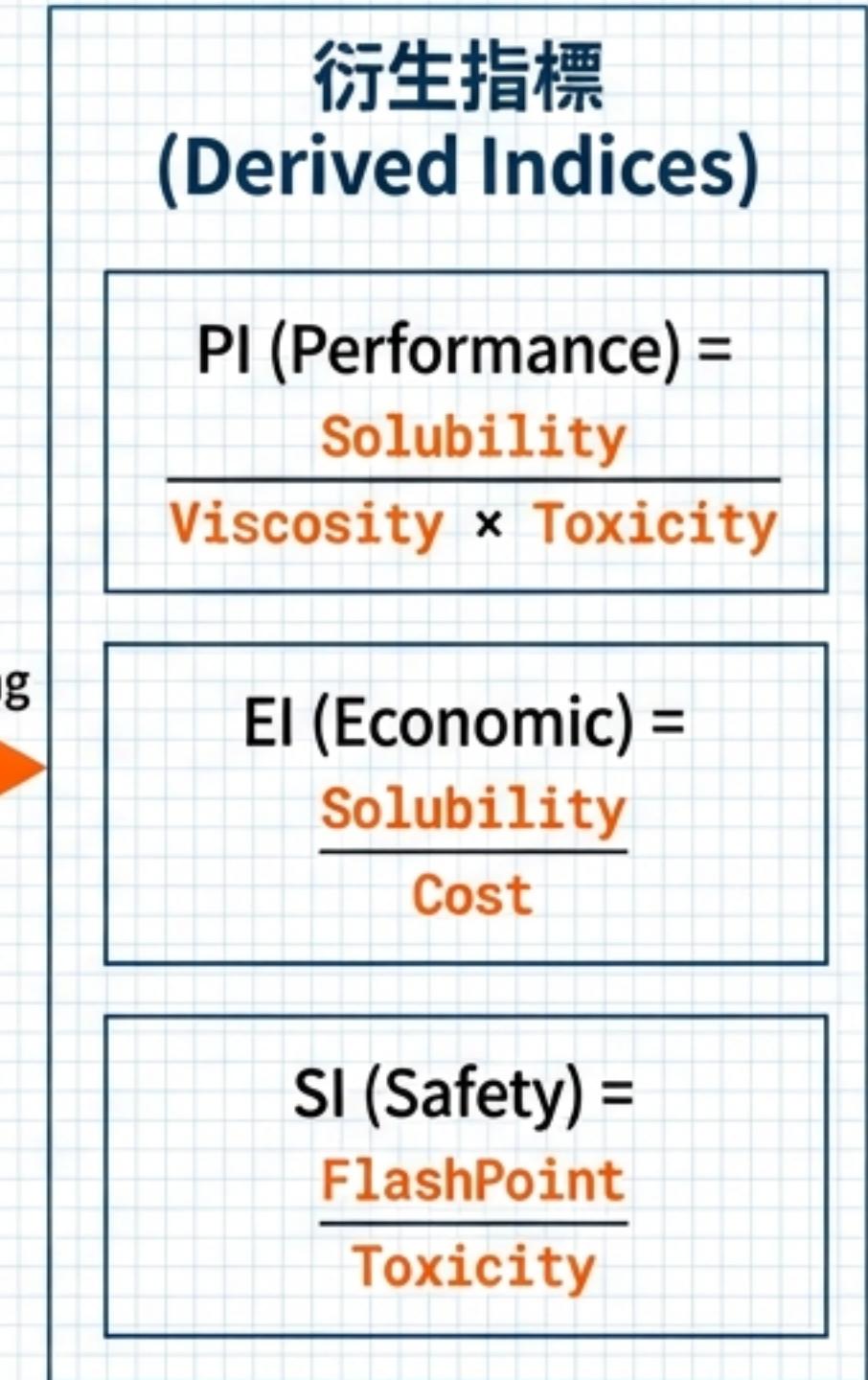
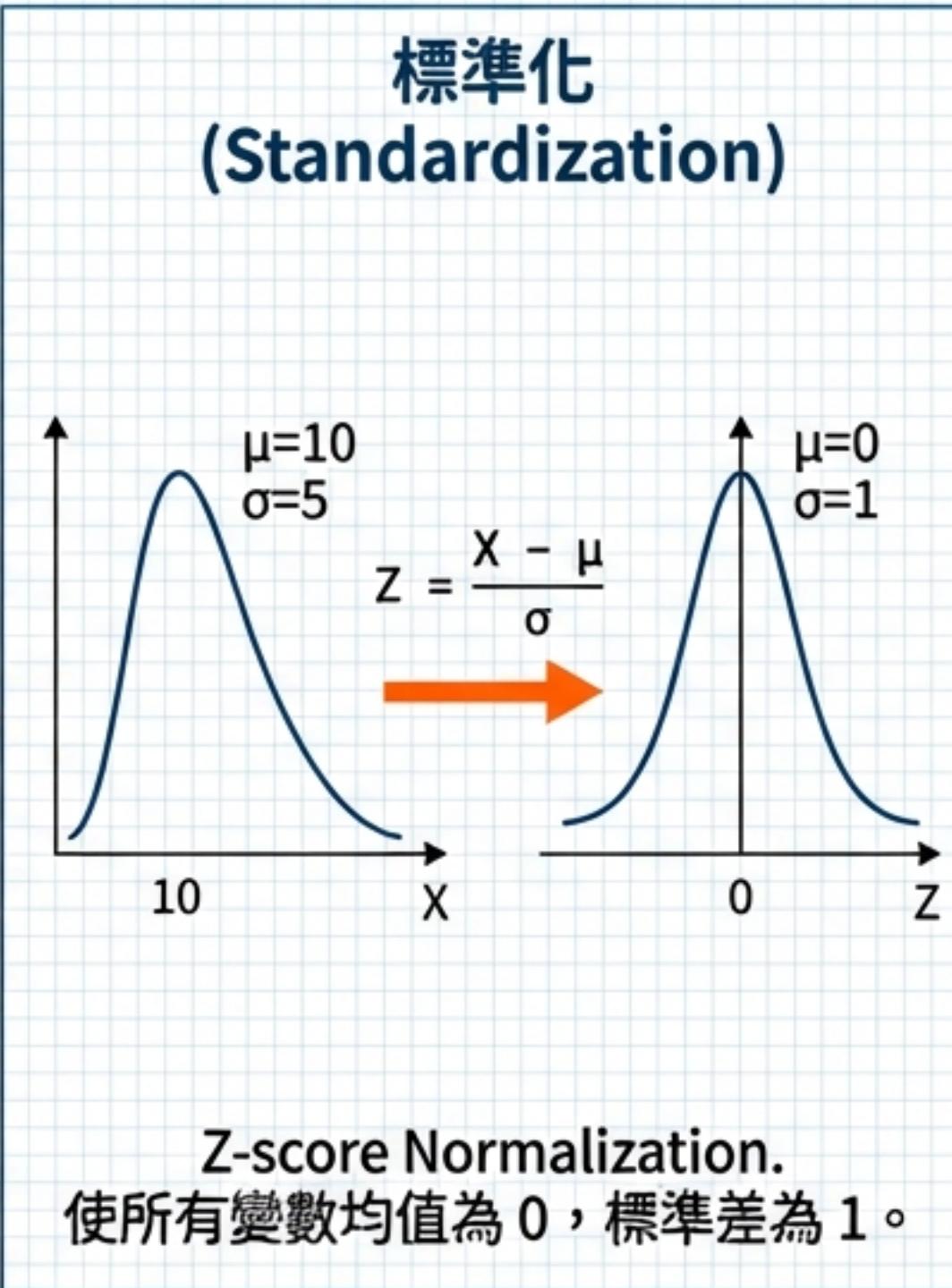


數據基礎與前處理：工欲善其事，必先利其器

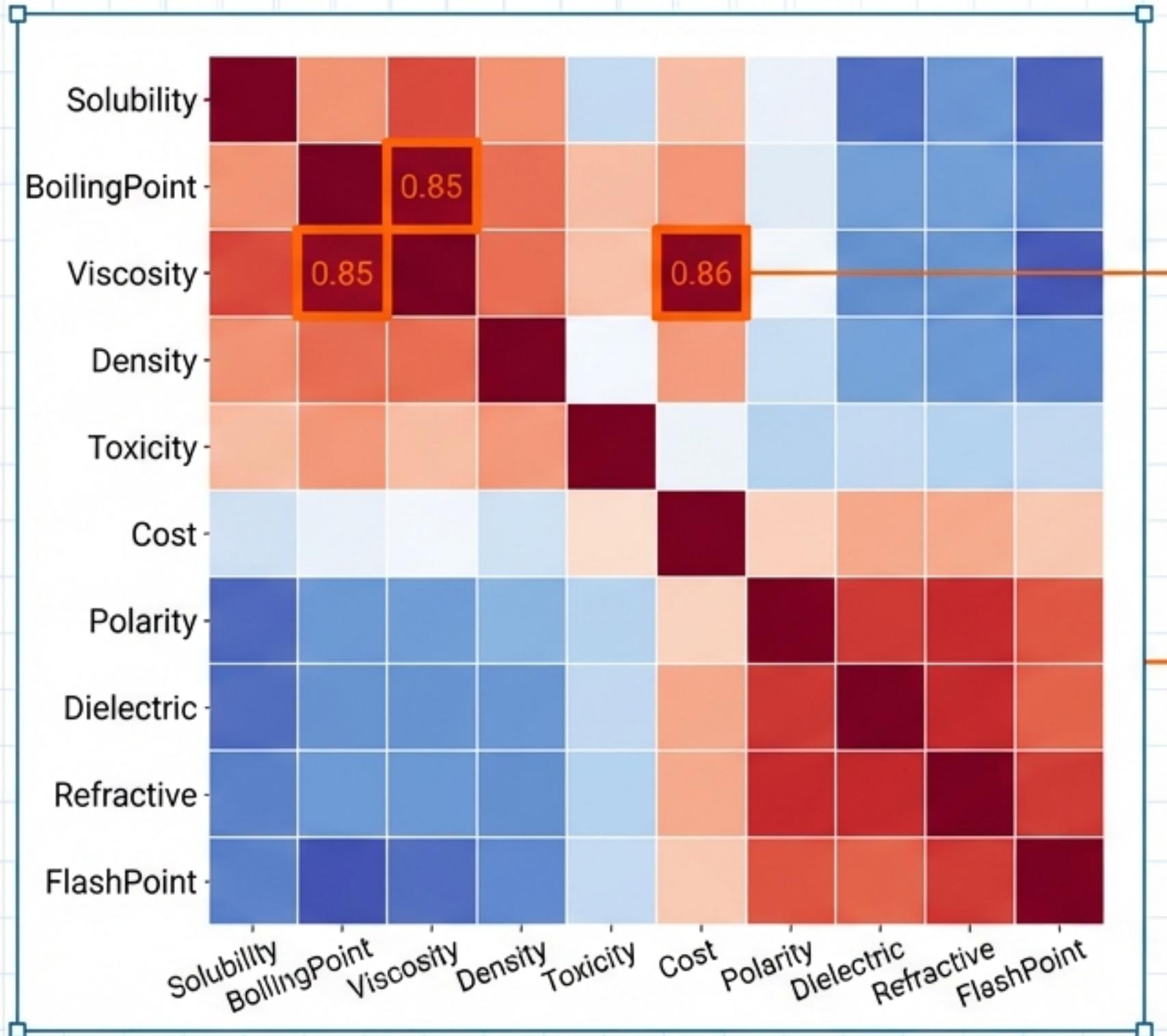
原始特徵 (Raw Features)

Variable	Example Value
Solubility	0.5
BoilingPoint	100
Viscosity	0.5
Density	19.75
Toxicity	0.38
Cost	200
Polarity	0.5
DielectricConstant	0.194
RefractiveIndex	0.001
FlashPoint	19.5

⚠ Scale Difference!
(e.g., BP ~100 vs Viscosity ~0.5)



數據會說話：變數間的相關性分析



成本驅動因素 (Cost Drivers)

Viscosity \leftrightarrow Cost ($r=0.86$)
重質溶劑通常更昂貴。

物理性質聯動 (Physical Link)

BoilingPoint \leftrightarrow Viscosity ($r=0.85$)。
高沸點通常伴隨高黏度。

多重共線性 (Multicollinearity)

變數間高度相關，這正是降維分析
(PCA) 的最佳應用場景。

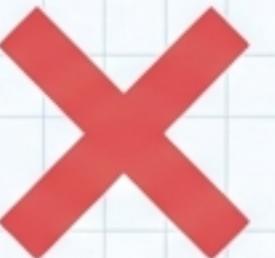
階段 1：物以類聚 (Clustering Analysis)

Method: K-Means (K=4)

Cluster 1: High Risk (Eliminate ✗)

Boiling Point: 179°C | Toxicity: High

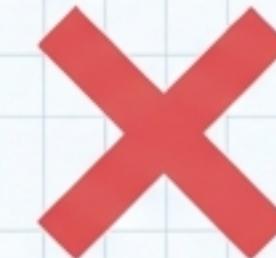
回收困難，毒性過高。



Cluster 2: High Polarity (Eliminate ✗)

Polarity: 8.1 | Dielectric: 25

與水互溶性過高，不利萃取。



Cluster 0: Economic (Keep ✓)

Cost: Low | Polarity: Low

經濟型選擇，適合成本敏感應用。



Cluster 3: Balanced (Keep ✓)

Viscosity: 0.54 | Toxicity: Low

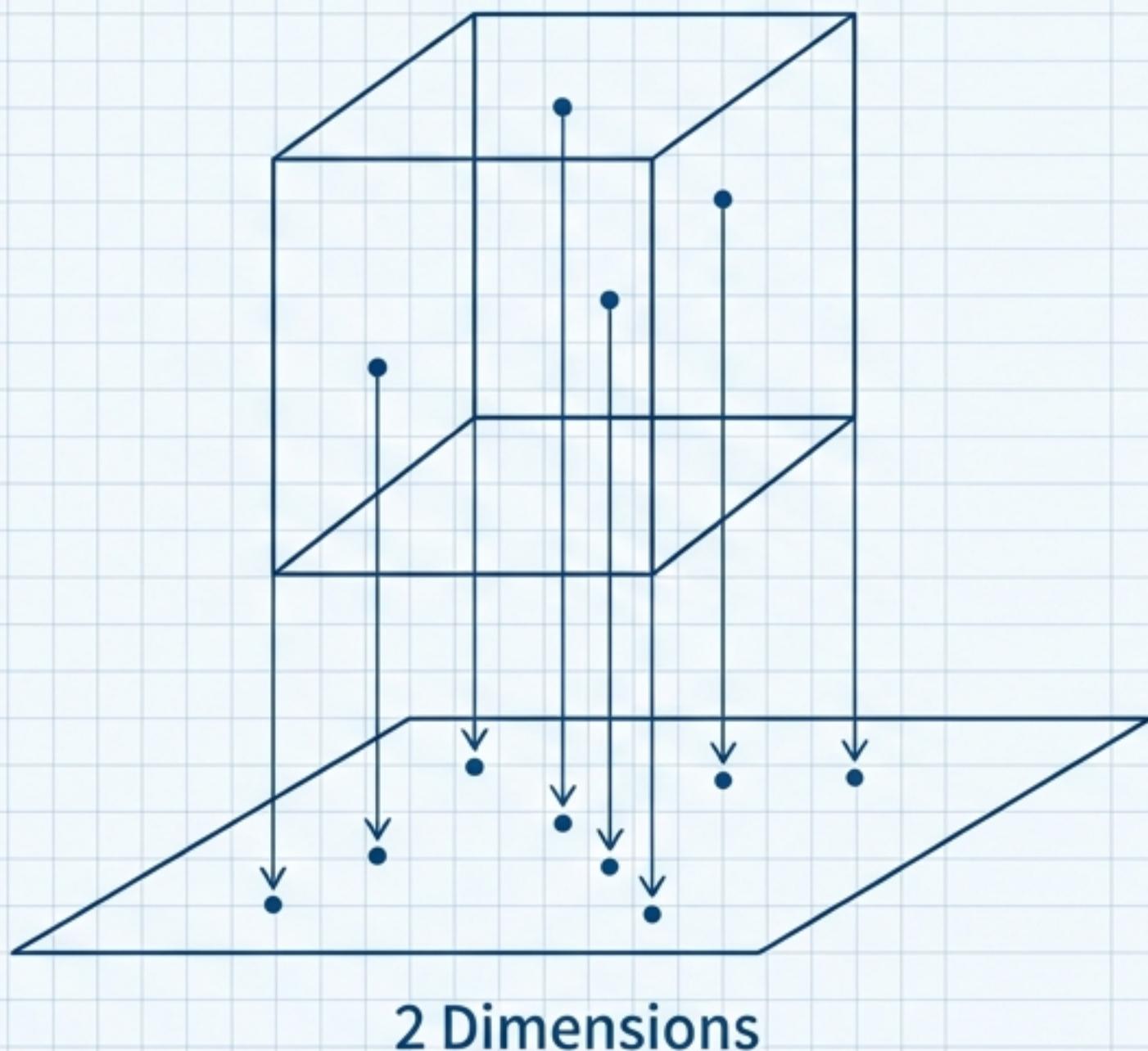
綜合性能最佳，優質候選。



Result: 100 Candidates → 50 Candidates (Cluster 0 & 3)

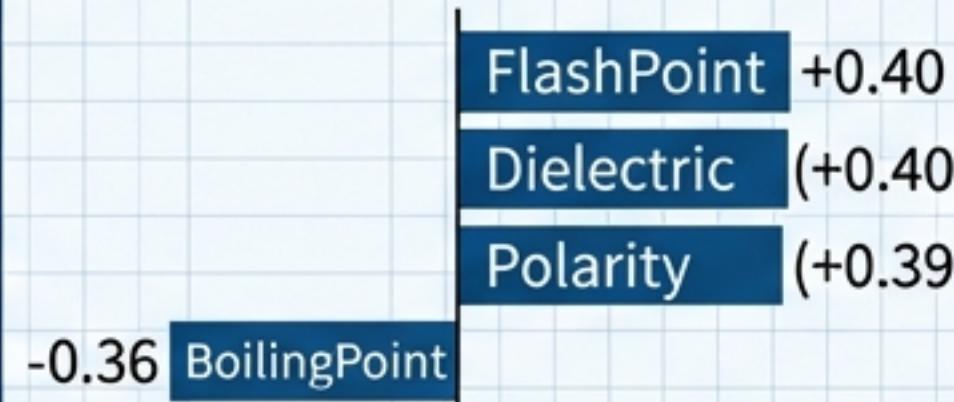
階段 2：降維原理與物理意義 (PCA Principles)

10 Dimensions



Loadings Analysis

PC1 - The Polarity Axis (42.2% Variance)



High PC1 =
高極性 / 低揮發性

PC2 - The Cost/Risk Axis (11.9% Variance)

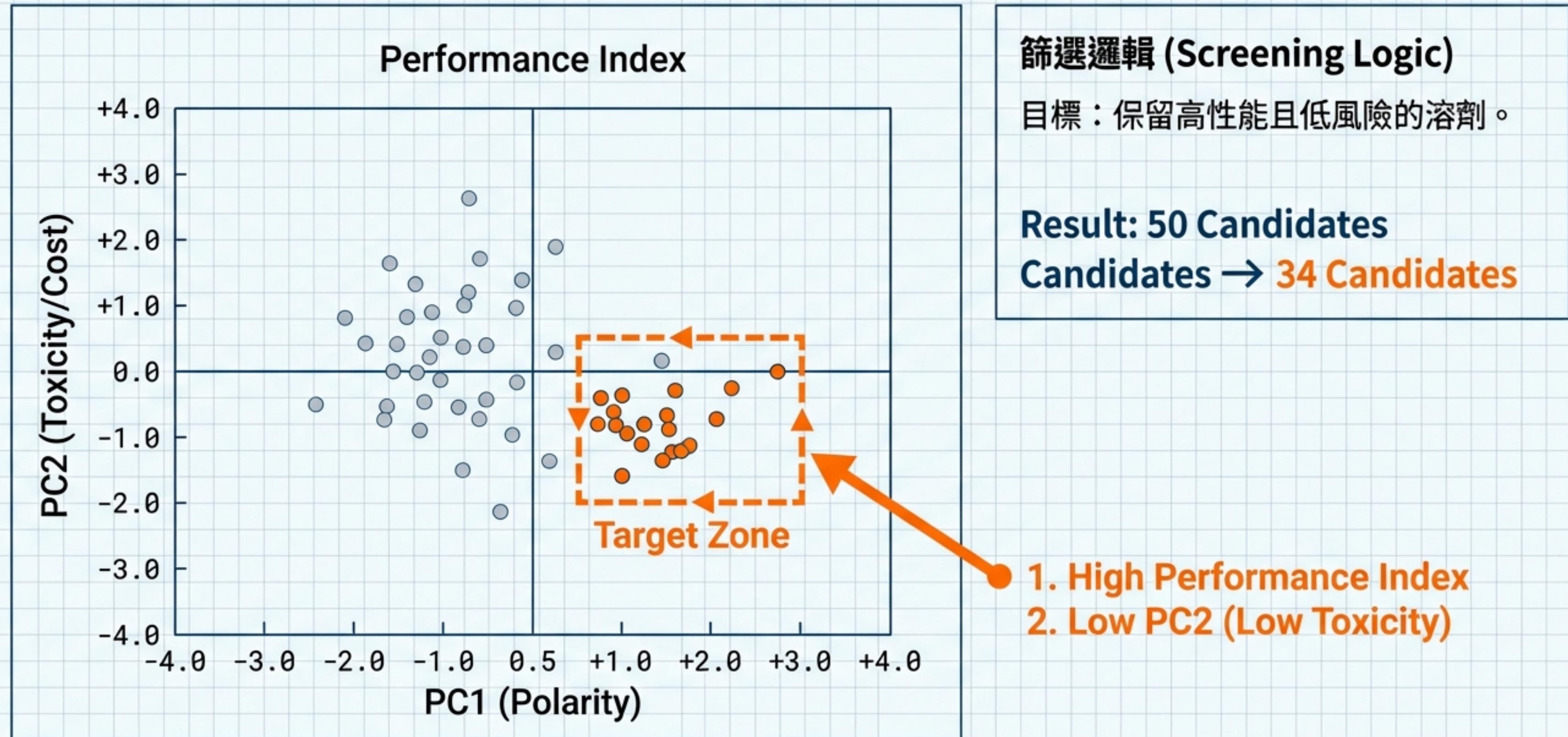


High PC2 =
高毒性 / 高成本

Key Insight Box

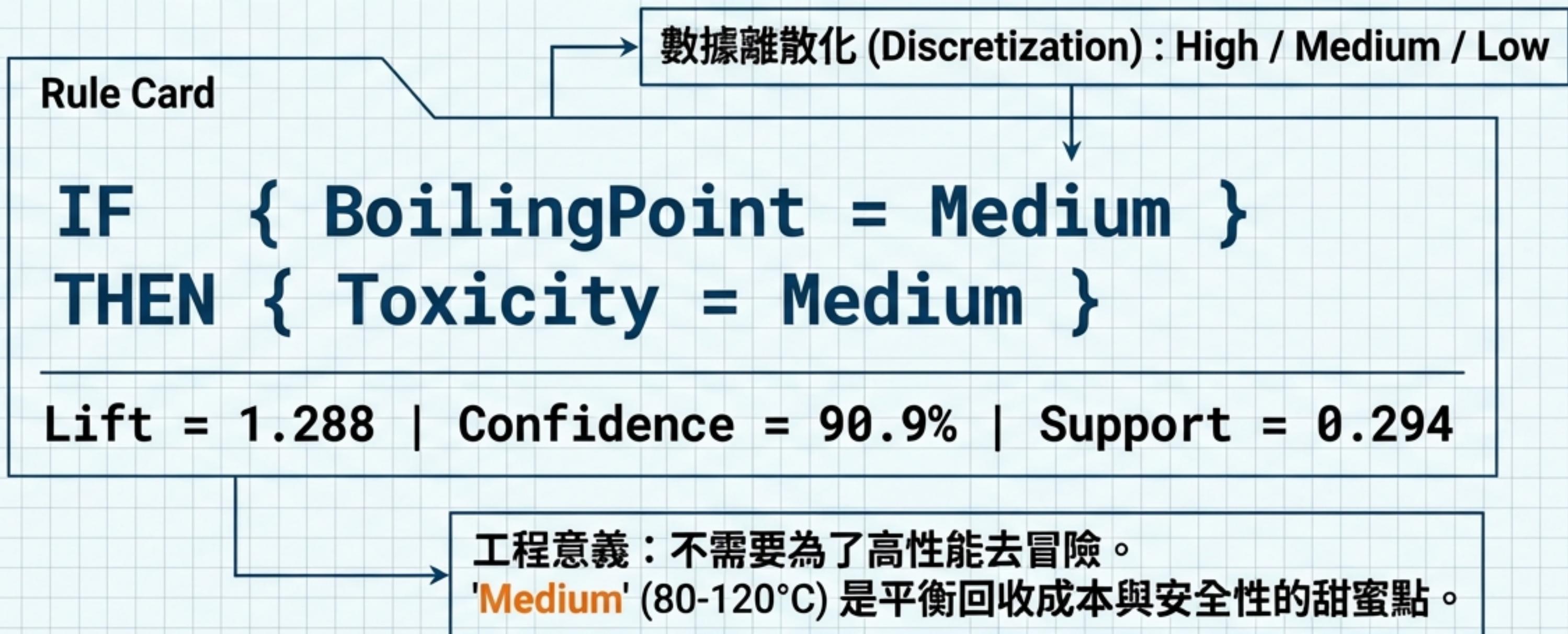
PCA 就像是一個指南針，PC1 指示極性方向，
PC2 指示風險方向。

階段 2：在 PC1-PC2 平面上尋找黃金區域



階段 3：挖掘隱藏邏輯 (Association Rules)

Method: Apriori Algorithm



Result: Validates strategy to target **balanced solvents**. Candidates refined to **Top 5**.

決策時刻：建立綜合評分標準 (Final Scoring Model)

$$\text{Composite Score} = 0.4 \times \text{PI} + 0.3 \times \text{EI} + 0.3 \times \text{SI}$$

40% Performance (PI)



效率至上 (Efficiency First).
Solubility / Viscosity.

30% Economy (EI)



經濟可行 (Viability).
Solubility / Cost.

30% Safety (SI)



安全無虞 (Safety).
FlashPoint / Toxicity.

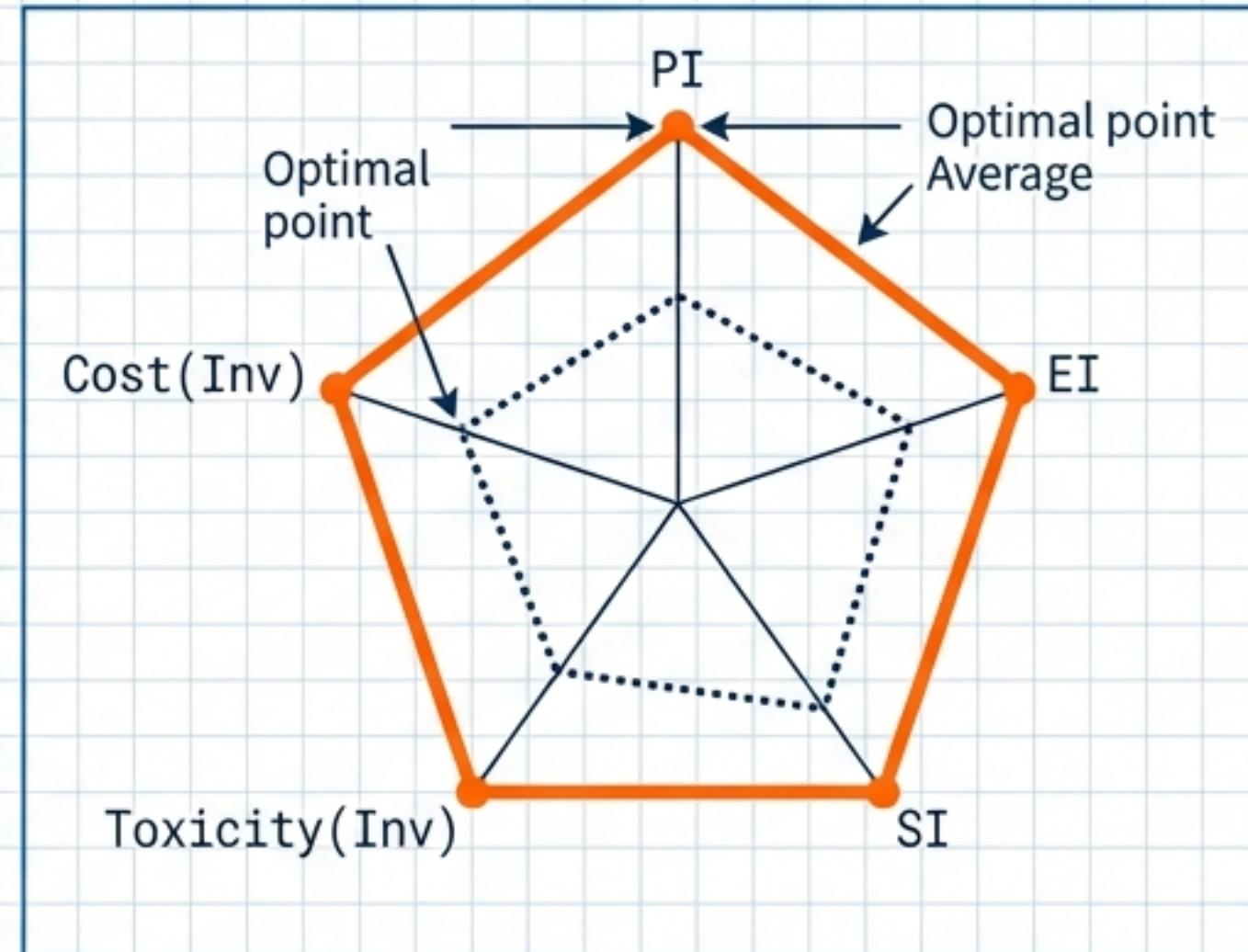
所有指標在計算前均經過歸一化處理 (Normalized).

最終結果：Top 5 最佳候選溶劑

Top 5 Candidates

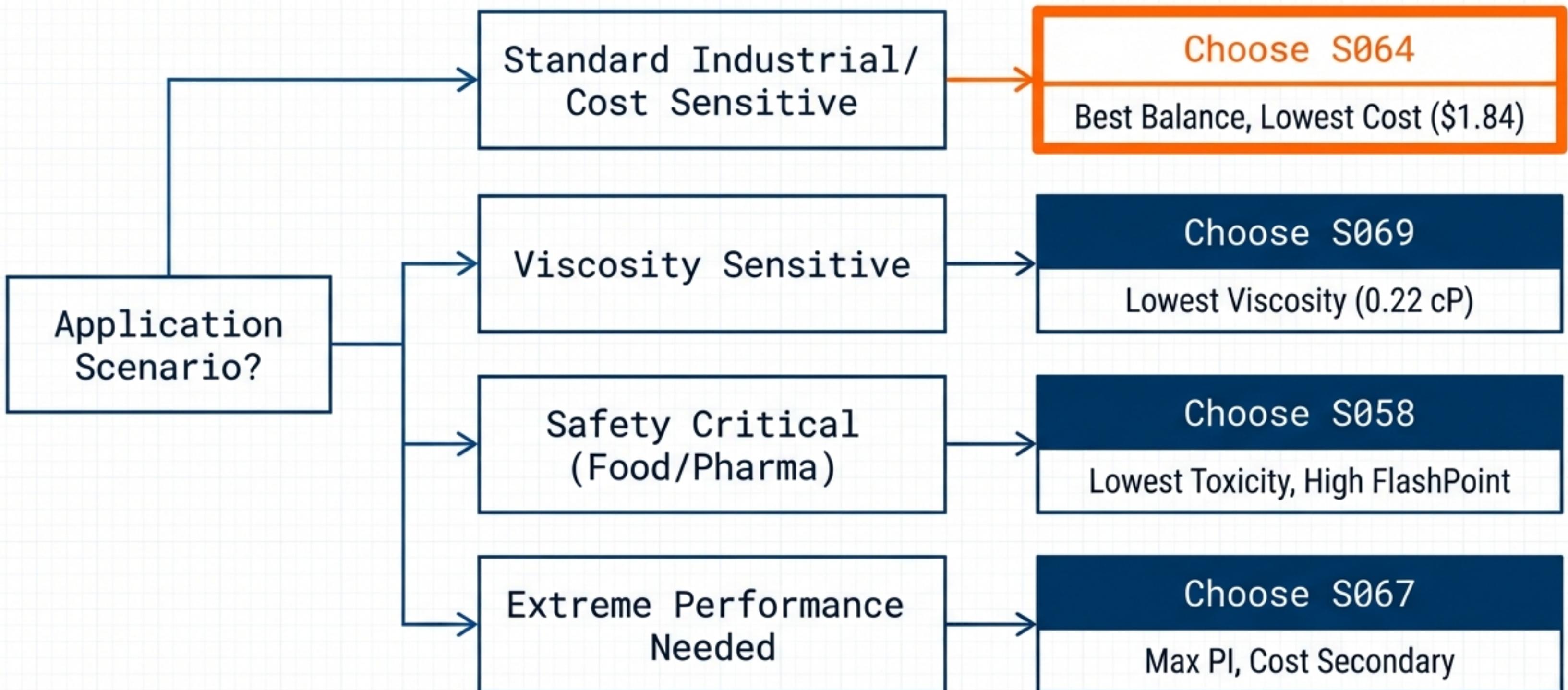
Rank	ID	Score	Toxicity	Cost	Note
1	S064	0.68	2 (Low)	\$1.84	Best Balance
2	S069	0.57	4 (Med)	\$1.65	Best PI
3	S058	0.57	2 (Low)	\$2.86	Safest
4	S067	0.47	4 (Med)	\$4.50	Max Perf.
5	S052	0.47	3 (Low)	\$2.11	High Sol.

Champion: S064



Score: 0.68 (19% higher than 2nd place). **No weaknesses.**

情境分析：沒有最好的溶劑，只有最適合的溶劑



方法論協同效應：為何需要組合拳？

Method	Strength	Blind Spot	Role in Funnel
Clustering	Groups similar items	Doesn't rank 'better' vs 'worse'	Filter Noise (Stage 1)
PCA	Visual Map of space	Loses ~45% variance info	Navigation (Stage 2)
Association Rules	Discovers Logic	Hard to rank specific candidates	Validation (Stage 3)

Combined Strength: Filter → Map → Validate → Rank

領域知識的價值：AI 與化工工程師的協作



工程師的角色 (The Engineer's Role)

- 定義指標 (Defining PI, EI, SI)
- 解釋物理意義 (Interpreting PC1 as Polarity)
- 權衡決策 (Trade-off: Cost vs Safety)

AI 的角色 (The AI's Role)

- 快速處理 (Processing 100 samples)
- 發現關聯 (Correlations: Viscosity ↔ Cost)
- 降維視覺化 (Mapping 10D to 2D)

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“AI 不會取代化工工程師，但懂得使用 AI 的工程師將取代不懂的人。”

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結論與下一步：從模擬走向實驗

Impact Summary

95% Reduction
Experimental Effort



Seconds vs Weeks
Screening Speed



Next Steps (Action Plan)

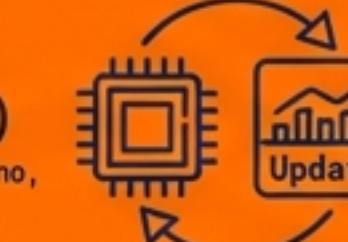
Lab Verification
(Test S064)

Noto Sans TC & Roboto Mono,
Deep Engineering Blue



Feedback Loop
(Update Model)

Noto Sans TC & Roboto Mono,
Deep Engineering Blue



Scale Up
(Pilot Test)

Noto Sans TC &
Roboto Mono,
Deep Engineering Blue

