

# Who Benefits from Scientific Entrepreneurship Training?

Diagnostic Results: Attrition, Multiple Outcomes, Theory Component — v6

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**Where we stand.** The main GenericML scanner (v5) tested 12 moderators across cognitive, venture, and belief-calibration dimensions on  $\Delta$ SI. All 12 were null ( $p > 0.20$ , 250 splits each). This note reports three diagnostic analyses run to understand whether heterogeneity exists along other dimensions before deciding on the paper’s framing.

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## D1 — Attrition deep-dive.

*What we did.* Using the full panel ( $N_{\text{baseline}} = 1,187$  across 3 arms), we computed retention rates by arm  $\times$  period, estimated the ATE trajectory (mean  $\Delta$ SI by arm and period), and ran a logit of attrition on arm dummies and site fixed effects for each follow-up period. We also compared baseline characteristics of stayers vs. droppers by arm.

*What we found.*

- **Retention by period 3:** Control 22.6% · T&E 44.4% · Evidence-Based 50.8%. The control arm loses 77% of its sample by period 3; treated arms lose 50–55%.
- **Differential attrition is large and highly significant.** T&E and Evidence-Based founders are  $\approx 32$  percentage points *less* likely to drop out than control at every follow-up period (logit AME  $\approx -0.32$ ,  $p < 10^{-10}$ ).
- **The ATE “fade-out” is not reversion—it is control-group recovery.** The control arm’s mean  $\Delta$ SI is *negative* in periods 1–3 ( $-0.34$ ,  $-0.39$ ,  $-0.17$ ) before recovering to near zero in periods 4–5. T&E is positive in periods 1–2 ( $+0.26$ ,  $+0.20$ ) and also near zero in periods 3–5. The effect disappears because the control group drifts back up, not because the treated group reverts.
- **Selective survival.** In treated arms, stayers are systematically older, more experienced, and more ambitious than droppers (e.g., `work_experience_year`  $p = 0.003$ ; `idea_aspiration_1`  $p < 0.001$  in EB). Control stayer/dropper differences are mostly non-significant.

*Implication.* Attrition is high and differential. Period 3+ comparisons are confounded by selection. The periods 1–2 results are the most credible estimates of the causal effect. **D2 (heterogeneous fade-out) is not viable.** Any reframe of the paper must be grounded in periods 1–2.

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## D3 — Multiple outcomes: does behavioral HTE exist where SI-HTE does not?

*What we did.* We re-ran the GenericML BLP scanner (250 splits) for three behavioral outcomes at period 1, using the top 4 moderators from the main scanner:

Outcome	Construct	ATE	ATE $p$
<code>delta_si</code>	Scientific Intensity change (reference)	+0.60	< 0.001
<code>pivoting_any</code>	Any pivot made (binary)	+0.17	0.004
<code>sales_dummy</code>	Made a sale (binary)	-0.03	0.46
<code>dropout_program</code>	Left the program (binary)	<i>no valid splits</i>	

*What we found.* All HTE coefficients ( $\hat{\beta}_2$ ) are null across all outcome  $\times$  moderator combinations ( $p > 0.17$  in all 12 cells). Notably:

- `pivoting_any`: the ATE is real and significant ( $+0.17$ ,  $p = 0.004$ )—the training increases pivoting on average—but this effect is uniform across founder types.
- `sales_dummy`: no ATE and no HTE. The training does not affect sales in period 1.
- `dropout_program`: zero valid splits, suggesting near-zero variance in program dropout at period 1 (almost everyone still enrolled). Variable may be better defined at a later period.

*Implication.* **Ruta 1 (different Y) is not viable.** Behavioral translation of SDM learning is also uniform across observable founder types.

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## D4 — T&E vs Evidence-Based: does the theory component have differential fit?

*What we did.* We redefined the treatment as T&E (arm 1) vs. Evidence-Based (arm 2), excluding the control arm. The difference between these two arms is precisely the SDM theory component. We ran the full 12-moderator GenericML scanner (250 splits, periods 1 and 2).

*What we found.* ATE (T&E – EB) is near zero and non-significant at both periods ( $\approx -0.02$  at P1,  $+0.07$  at P2). All 24 HTE coefficients ( $\hat{\beta}_2$ ) are null ( $p > 0.15$  in all cells). The closest to signal is `exit_probability` at period 2 ( $p = 0.148$ ), which does not meet any threshold.

*Implication.* T&E and Evidence-Based produce statistically indistinguishable outcomes on  $\Delta SI$ , for all founder types. **Ruta 3 (theory-component HTE) is not viable.** The SDM framework does not differentially benefit any identifiable founder type relative to evidence-only training.

**Synthesis: where we are.**

Route	What it needed	Status	Evidence
<b>0</b> — <b>Original</b>	Cognitive battery signal (Diego)	<i>Pending</i>	Awaiting 17 items
<b>1</b> — <b>Different Y</b>	HTE on pivot / sales / dropout	<b>Closed (D3)</b>	All null
<b>2</b> — <b>Fade-out</b>	Genuine fade-out (D1) + heterogeneous (D2)	<b>Closed (D1)</b>	Attrition artifact
<b>3</b> — <b>Theory fit</b>	T&E vs EB HTE on any moderator	<b>Closed (D4)</b>	All null
<b>4</b> — <b>Methodological</b>	What we already have	<b>Open</b>	RF $\neq$ DML; CLAN portrait

**The paper has one open route before Diego:** Ruta 4. The IoC finding (RF-predictive  $\neq$  causally significant) combined with the CLAN portrait of the high-benefit founder is a clean methodological contribution to how we use ML for program targeting. It is publishable without the cognitive battery.

**If Diego delivers the 17 items** (risk aversion, uncertainty aversion, learning orientation, competitiveness), Ruta 0 reopens: these are the most theory-grounded moderators and the only untested family. A signal there makes this a complete foundations paper.

**Scripts:** 05\_attrition\_dynamics.do (D1) · 06\_genericml\_d3\_multiy.R (D3) · 07\_genericml\_d4\_te\_vs\_eb.R (D4)  
**Data:** ERC\_RED\_HMC\_clean.dta ( $N = 6,732$  obs, 6 sites, 3 arms)    **Method:** Chernozhukov et al. (2025, *Econometrica* 93(4))  
 All results labeled EXPLORATORY.