

Key Take Aways: (1) Aβ-positive late-onset Amnestic MCI is not a uniform phenotypic entity; 6 neuropsychologically distinct phenotypes were identified in ADNI that have a characteristic cognitive profile and rate of decline. (2) The Amnestic-predominant phenotype showed a greater rate of clinical decline and cortical atrophy in the medial temporal lobes compared to the Non-Amnestic phenotypes (Dysexecutive, Language, Visuospatial). (3) Cognitive phenotyping at the MCI stage could offer a foundation for more precise patient stratification in clinical trials, person-specific prognostic counseling, and personalized disease management.

BACKGROUND

• Late-onset AD (LOAD) is traditionally conceptualized as a clinically homogeneous **amnestic** syndrome, yet accumulating evidence suggests substantial heterogeneity in its cognitive and neurobiological features^{1,2,3}.

• We do not yet understand:

- If this phenotypic heterogeneity is present at the earlier stage of amyloid-positive late-onset amnestic mild cognitive impairment (LOAD-aMCI)
- If the heterogeneity in LOAD-aMCI is consistent with what has been observed in early-onset AD (EOAD-aMCI)^{4,5}.
- The downstream implications of phenotypic heterogeneity, including different rates of decline and regional differences in brain structure.

• The goals of the current study are:

1. Characterize phenotypic heterogeneity in the ADNI biomarker-confirmed late-onset aMCI sample using rule-based cognitive subtyping.
2. Determine how these phenotypes differ on (a) Baseline demographic and clinical characteristics, (b) Longitudinal trajectory of clinical decline, and (c) baseline structural atrophy patterns.

• Characterizing phenotypic heterogeneity within Aβ-positive LOAD-aMCI has direct implications for clinical trial design, prognosis, and personalized care. Pooling phenotypically distinct patients may dilute treatment effects and obscure differential progression rates

METHODS

• We studied **390 Aβ-positive MCI** with symptom onset after the age of 65 (LOAD-aMCI) from the Alzheimer's Disease Neuroimaging Initiative (ADNI). Data was pooled across ADNI-1, ADNI-Go, ADNI-2, and ADNI-3.

• All had complete baseline neuropsychological assessment and longitudinal ratings on the Clinical Dementia Rating scale (CDR). A subset (N=306) had longitudinal Montreal Cognitive Assessment (MoCA) scores. Most (N= 387) had baseline cortical thickness MRI data.

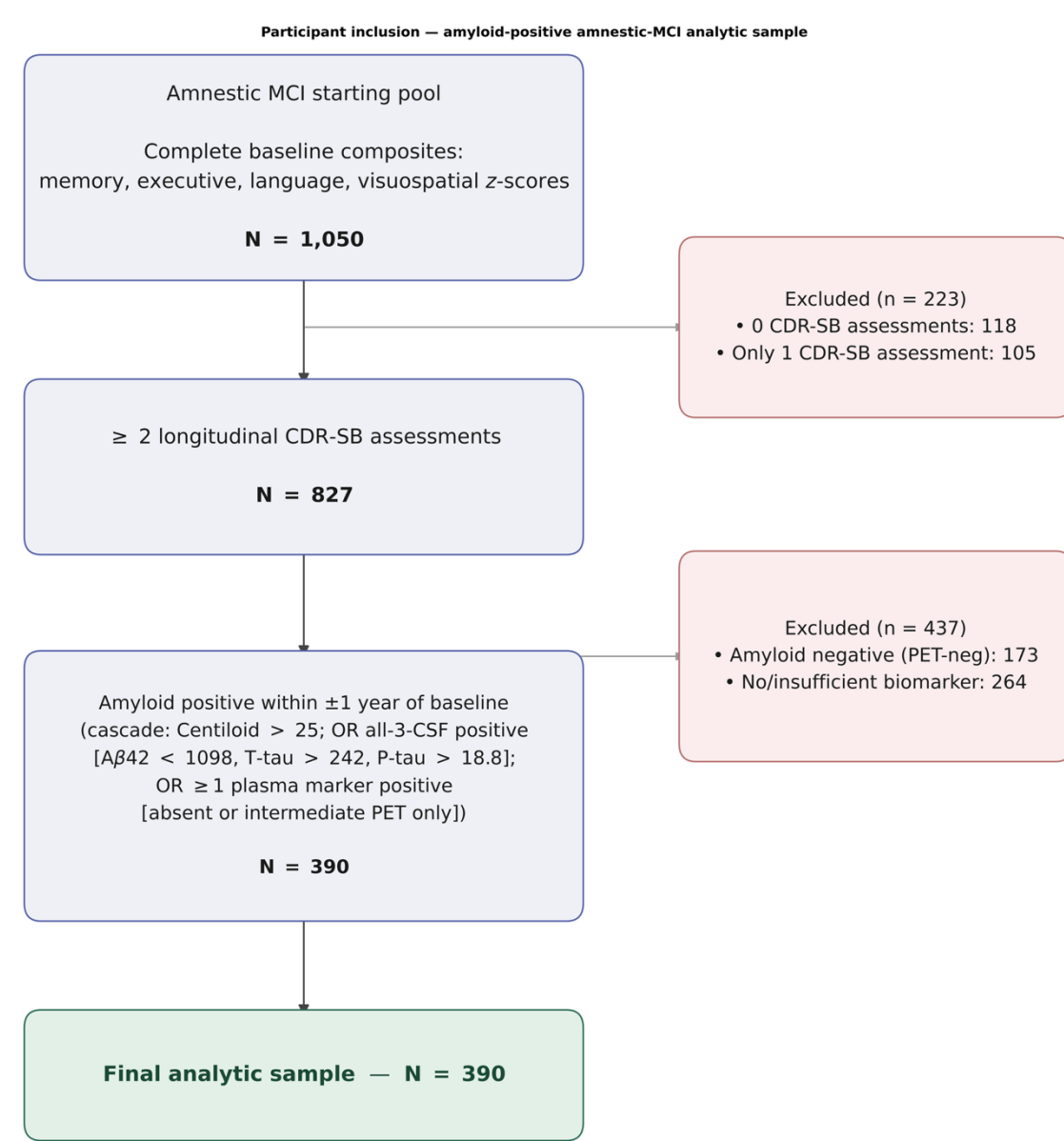
• **Person-centered rule-based cognitive phenotyping** was applied, consistent with prior phenotypic subtyping approaches^{1,2,5}. See flowchart in Results.

• **Clinical trajectory modeling:**

- Linear mixed models (LMM) with random intercepts and random slopes were used to estimate longitudinal CDR-SB trajectories. Models included phenotype × time interactions and controlled for baseline CDR-SB.

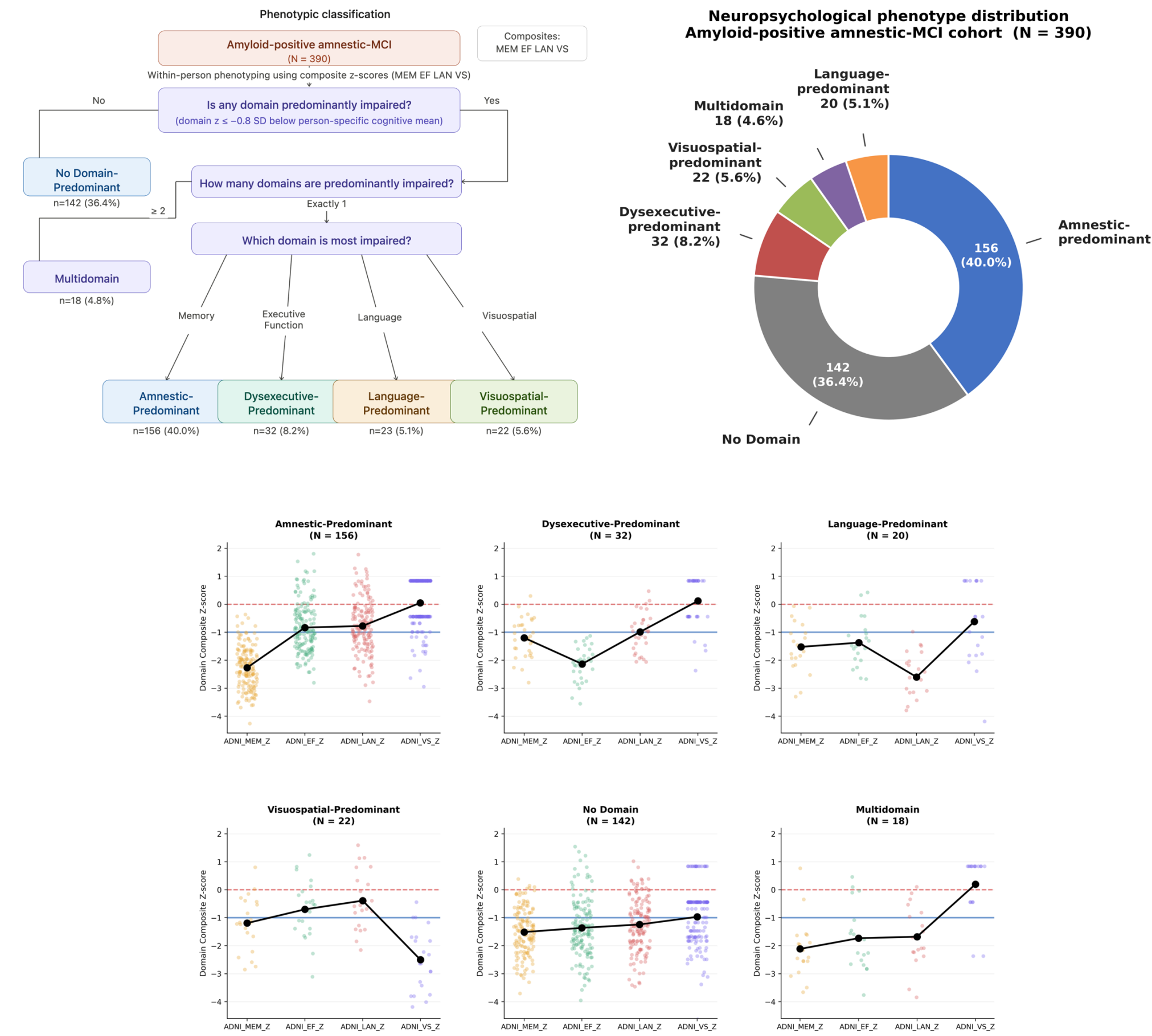
• The same LMM approach was used to estimate longitudinal MoCA trajectories.

• **Structural MRI:** Vertex-wise GLM were computed across the cortical surface using FreeSurfer to identify regions of significant atrophy, covaried for age and sex. Cluster-based correction for multiple comparisons was applied (p < .05, two-tailed). Mean cortical thickness was extracted from anatomically defined ROIs. One-way ANOVAs tested group differences across the neuropsychological phenotypes, with post-hoc pairwise comparisons (Tukey HSD) and a false discovery rate correction (q < .05).



RESULTS

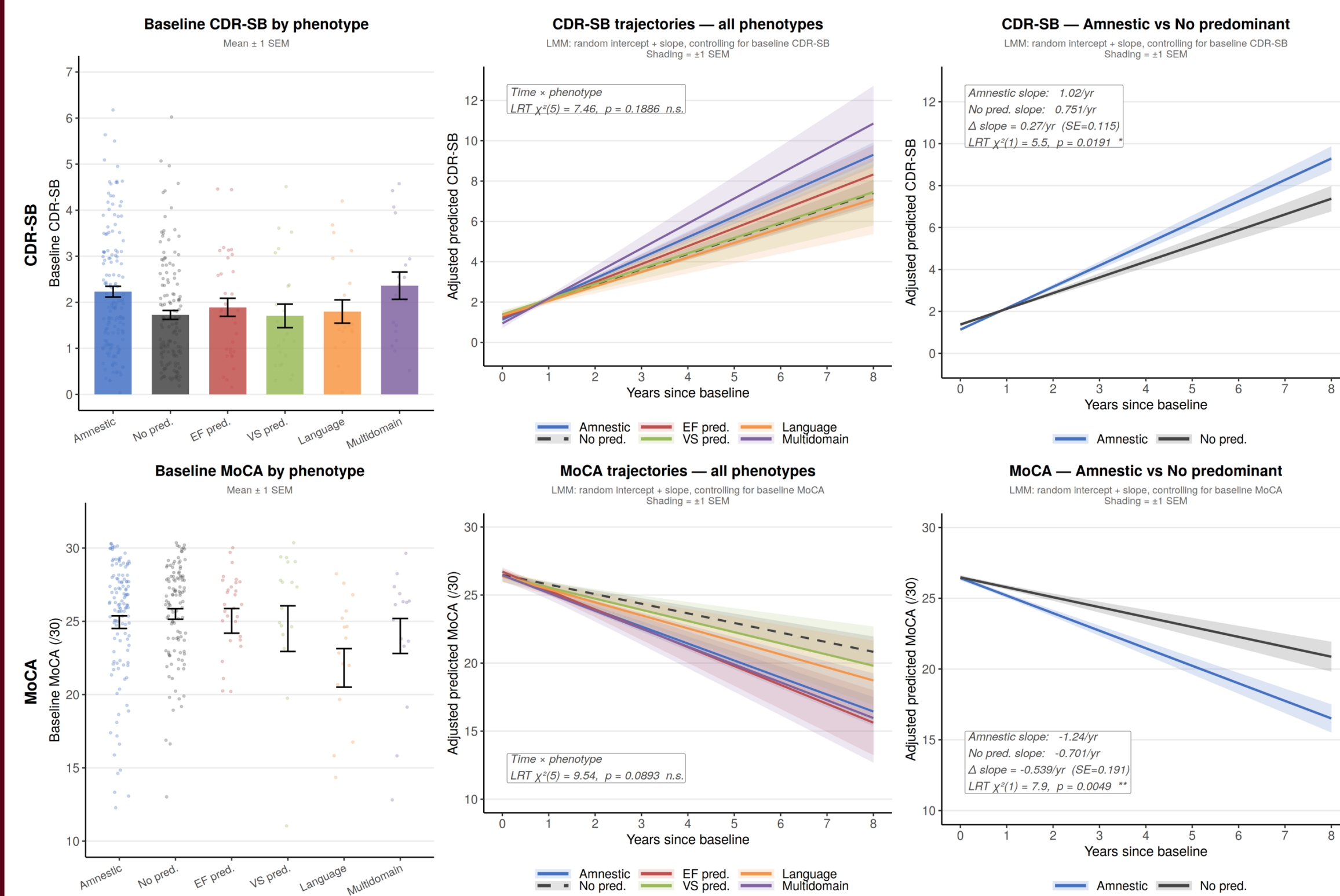
6 Distinct Cognitive Phenotypes



	Amnestic-predominant	Dysexecutive-predominant	Language-predominant	Visuospatial-predominant	Multidomain	No Domain	Total	p value
n	156	32	20	22	18	142	390	
Age, years	72.3 (6.6)	76.1 (6.3)	73.5 (5.0)	72.1 (7.4)	75.4 (8.7)	74.3 (7.0)	73.5 (6.9)	0.020^a
Female sex	60 (38.5%)	13 (40.6%)	16 (80.0%)	6 (27.3%)	6 (33.3%)	68 (47.9%)	169 (43.3%)	0.005^b
Education, years	16.2 (2.6)	16.9 (2.5)	15.1 (2.5)	16.3 (2.6)	15.3 (3.1)	15.6 (3.0)	15.9 (2.8)	0.053 ^a
CDR Global	0.5 (0.5–0.5)	0.5 (0.5–0.5)	0.5 (0.5–0.5)	0.5 (0.5–0.5)	0.5 (0.5–0.5)	0.5 (0.5–0.5)	0.5 (0.5–0.5)	0.269 ^c
CDR-SB	2.0 (1.0–3.0)	2.0 (1.0–2.6)	1.5 (1.4–2.6)	1.2 (0.6–2.5)	2.2 (1.5–3.0)	1.5 (1.0–2.5)	1.5 (1.0–3.0)	0.016^c
MoCA	24.9 (4.8)	25.0 (4.6)	21.8 (5.4)	24.5 (6.6)	24.0 (4.6)	25.5 (3.8)	24.9 (4.6)	0.072 ^a
APOE ε4 homozygotes	35 (22.4%)	4 (12.5%)	2 (10.0%)	4 (18.2%)	1 (5.6%)	19 (13.4%)	65 (16.7%)	0.185 ^b

^aOne-way ANOVA. ^bChi-square. ^cStudent's t-test. Bold = most extreme group mean/percentage when p < 0.05.

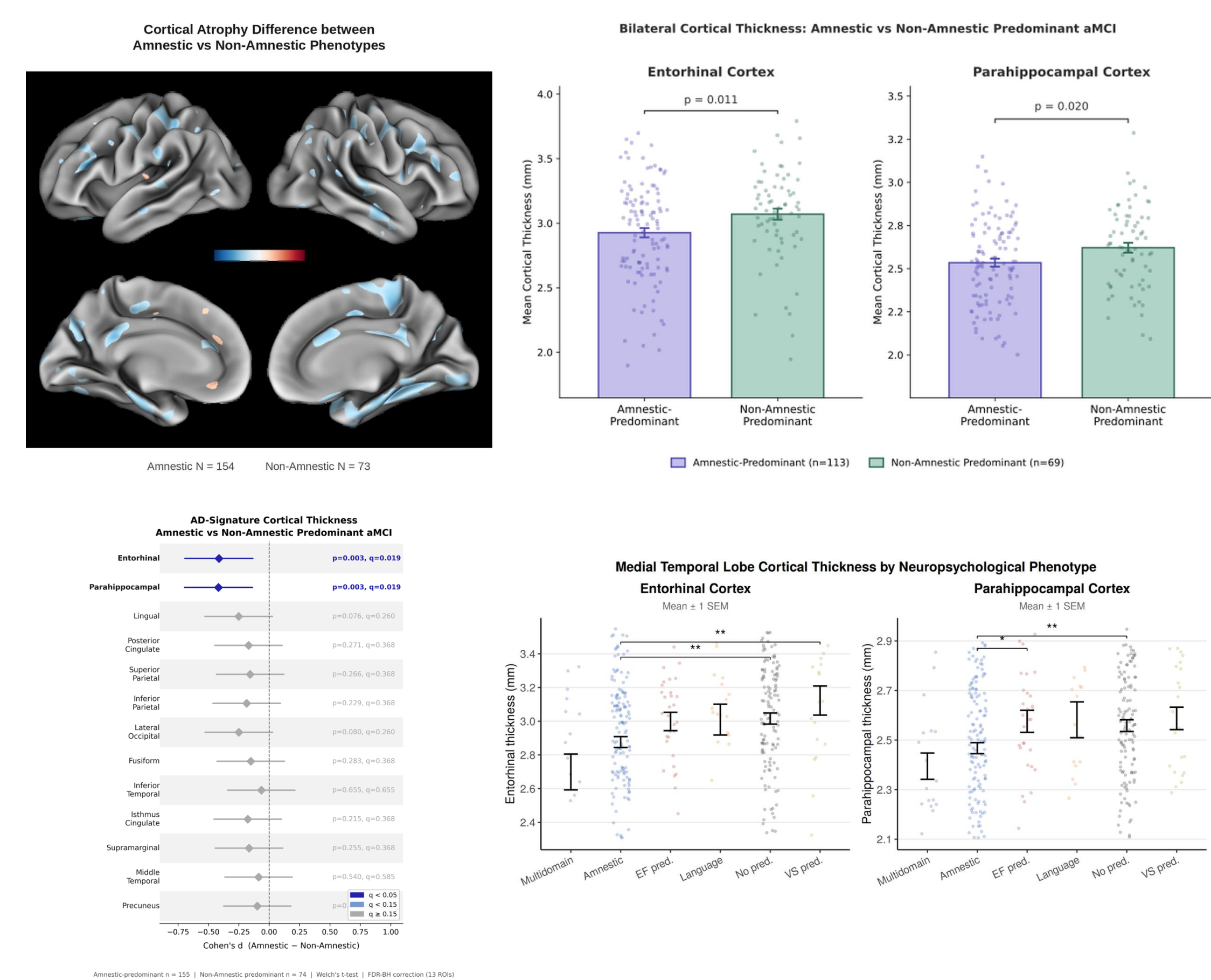
Phenotypes differ on longitudinal trajectories of clinical decline (CDR-SB and MoCA)



Baseline pure Amnestic or Multidomain phenotypes decline more quickly over time functionally (CDR-SB) and cognitively (MoCA) compared to Non-Amnestic or No-predominant phenotypes

RESULTS

Phenotypes differ on extent of MTL atrophy



The Amnestic-predominant and Multidomain phenotypes have greater medial temporal lobe (MTL) cortical atrophy compared to the Non-Amnestic presentations

CONCLUSIONS

• Six distinct Aβ-positive late-onset aMCI cognitive phenotypes were identified with cognitive rule-based phenotyping.

• **Longitudinal CDR-SB and MoCA trajectories diverge between phenotypes**, such that pure Amnestic and Multidomain baseline phenotypes decline fastest, while the group with no predominant domain of impairment declines slowest.

• **Structural neuroimaging supports the validity of phenotypic classification**, with greater MTL atrophy in Amnestic vs Non-Amnestic Predominant phenotypes (Dysexecutive, Language, Visuospatial).

• These findings support a **phenotype-informed framework for studying and caring for amyloid-positive MCI**, in which the neuropsychological profile, rate of clinical and cognitive decline, and cortical atrophy pattern each provide complementary information about disease expression.

• Future work should examine whether phenotype at the MCI stage predicts downstream conversion to distinct Alzheimer's dementia subtypes (e.g., PCA, IvPPA, Dysexecutive Syndromes). Outcomes should extend beyond the memory-centric CDR-SB and MoCA screening tools.

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