

First-Episode Psychosis Patients Who Deteriorated in the Premorbid Period Do Not Have Higher Polygenic Risk Scores Than Others: A Cluster Analysis of EU-GEI Data

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Cluster studies identified a subgroup of patients with psychosis whose premorbid adjustment deteriorates before the onset, which may reflect variation in genetic influence. However, other studies reported a complex relationship

between distinctive patterns of cannabis use and cognitive and premorbid impairment that is worthy of consideration. We examined whether: (1) premorbid social functioning (PSF) and premorbid academic functioning (PAF) in

childhood and adolescence and current intellectual quotient (IQ) define different clusters in 802 first-episode of psychosis (FEP) patients; resulting clusters vary in (2) polygenic risk scores (PRSs) for schizophrenia (SCZ_PRS), bipolar disorder (BD_PRS), major depression (MD_PRS), and IQ (IQ_PRS), and (3) patterns of cannabis use, compared to 1,263 population-based controls. Four transdiagnostic clusters emerged (BIC = 2268.5): (1) high-cognitive-functioning ($n = 205$), with the highest IQ (Mean = 106.1, 95% CI: 104.3, 107.9) and PAF, but low PSF. (2) Low-cognitive-functioning ($n = 223$), with the lowest IQ (Mean = 73.9, 95% CI: 72.2, 75.7) and PAF, but normal PSF. (3) Intermediate ($n = 224$) (Mean IQ = 80.8, 95% CI: 79.1, 82.5) with low-improving PAF and PSF. (4) Deteriorating ($n = 150$) (Mean IQ = 80.6, 95% CI: 78.5, 82.7), with normal-deteriorating PAF and PSF. The PRSs explained 7.9% of between-group membership. FEP had higher SCZ_PRS than controls [$F(4,1319) = 20.4$, $P < .001$]. Among the clusters, the deteriorating group had lower SCZ_PRS and was likelier to have used high-potency cannabis daily. Patients with FEP clustered according to their premorbid and cognitive abilities. Pronounced premorbid deterioration was not typical of most FEP, including those more strongly predisposed to schizophrenia, but appeared in a cluster with a history of high-potency cannabis use.

Key words: premorbid/schizophrenia/cannabis/IQ/deterioration/bipolar

Introduction

Cognitive impairment is present in most patients with nonaffective psychosis, and a proportion of those with affective psychosis.^{1,2} The cognitive impairment in psychosis has been recently approached using clustering strategies.³ Research findings differ with some studies suggesting three cognitive clusters, among patients with nonaffective psychosis: those with high, intermediate, and low cognition,^{3,4} whereas others, including affective psychoses identified four.⁵ Cognitive clusters were also associated with symptoms and functional outcomes, indicating the utility of such classification for prognosis and treatment.^{6,7} Cluster analysis has also identified three cognitive-developmental trajectories from premorbid to current IQ: (a) stable impairment from childhood to current IQ, (b) normal premorbid function and high current IQ, and (c) IQ deteriorating from an average premorbid level.^{8–11}

Studies exploring premorbid adjustment using the Premorbid Adjustment Scale (PAS) replicated similar clusters using it as a composite score, including social and academic premorbid adjustment.^{12,13} However, when the social and academic scales were divided into two scores, six clusters were obtained.^{14,15} These findings suggest different premorbid social and academic trajectories,^{16,17} and

probably reflect differential development from childhood to adolescence. However, few studies examined patients with their first-episode of psychosis (FEP),^{6,18,19} and to the best of our knowledge, no studies clustered subjects according to their premorbid histories while also considering current IQ.

Other studies reported that patients with psychosis who were daily cannabis users constituted a less-impaired group who performed better in premorbid IQ^{20,21} and social, adjustment but had lower premorbid academic adjustment.²² Previous evidence revealed no significant differences in genetic load for schizophrenia between cannabis users and nonusers and between subjects smoking high-potency cannabis and other forms.^{23,24} Thus, we proposed that in such patients, deteriorating premorbid adjustment, when present, was due to substance abuse rather than a higher genetic susceptibility to schizophrenia.²²

Recent research examined the potential of multiple PRSs for schizophrenia, IQ, educational attainment, and attention deficit hyperactivity disorder (ADHD) in differentiating cognitive antecedents and current IQ clusters in a group of patients with established schizophrenia. The results suggested that different trajectories of cognitive development in psychosis may reflect variation in genetic influence.¹¹

The present study aimed to replicate these findings in a large, mostly European sample of FEP, compared with a group of population-based controls, cognitively characterized and profiled according to premorbid social and academic adjustment in childhood and early adolescence. In addition, we examined the discriminative ability of PRSs for schizophrenia (SCZ_PRS), bipolar disorders (BD_PRS), major depression (MD_PRS), and IQ (IQ_PRS) about cognitive profiles, as previous evidence in our sample suggests that different PRS differentiate between affective and nonaffective psychosis.²⁵ The principal aims were: (1) to examine whether participants with FEP can cluster in different groups according to measures of premorbid social functioning (PSF) and premorbid academic functioning (PAF) in childhood and early adolescence and current IQ; (2) to describe and compare the different clusters by PRSs for SCZ, BD, MD, and IQ. As a secondary aim, we sought (3) to examine whether patterns of cannabis use differ between clusters.

Methods

Participants

The sample comprised patients with FEP and population-based controls who participated in the multi-centre EU-GEI study; all the participants provided written informed consent. In addition, ethics committees in each study site provided ethical approval for the study.^{26–31} Details are in [Supplementary Material](#). Measures

Instruments. To facilitate replicability, we used measures of development and IQ, which are well-established and widely used in research. Namely, an abbreviated version of the Wechsler Adult Intelligence Scale (WAIS)³² and the Premorbid Adjustment Scale (PAS).^{33,34} The Cannabis Experience Questionnaire (CEQ) collected cannabis and other substance use information.²⁴ Researchers assessed patients as soon as they reached a stable mental state following treatment and ensured they were referring to the preonset period.²⁷ The CEQ and the PAS interviews were completed by at least one corroborative source of information (eg, family, clinical notes, and other clinicians) to minimize the effect of the recall bias. Details are in [Supplementary Material](#).

Cluster Analysis on FEP. We performed cluster analysis on childhood and early adolescence PSF, PAF, and current IQ to determine whether participants with FEP could be allocated into distinct membership classes. Controls were not included but used as a single group. We used a TwoStep Cluster Analysis procedure in SPSS, version 24 (details in [Supplementary Material](#)). We used the stepwise decrease in log-likelihood as the distance measure for identifying clusters and changes in the Bayesian Information Criterion (BIC) to determine the number of clusters to retain (best ratio change of cluster distance at least > 1.15).³⁵ Fleiss's kappa index established the extent of agreement in cluster assignment.³⁶ We inserted the variable "self-ascribed ethnicity" in the "evaluation" field of the two-step cluster analysis to estimate its descriptive importance comparing clusters. To see if PSF and PAF changes in the two age ranges were significantly different within the formed clusters and controls, we ran repeated-measures ANOVAs. It was adjusted by age, sex, country, and self-ascribed ethnicity, having clusters of patients and controls as the between-group factor.

Genotyping and PRS Calculation. Genotype procedure and genetic ancestry analysis are reported in [Supplementary Material](#). PRSs were built with the clumping and thresholding method using the PRSice software³⁷ and summary statistics from the largest available GWAS at the time of the analysis, excluding the current EU-GEI sample.³⁸⁻⁴¹ A *P*-value threshold of 0.05 for SNP inclusion was chosen across phenotypes based on the variance explained in the original studies. The number of SNPs overlapping with the discovery summary was 11,022 for the SCZ_PRS, 9,564 for the BD_PRS, 11,014 for the MDD_PRS, and 13,691 for the IQ_PRS. The estimated heritability indexed by common alleles explains 7% of the phenotypic variance for SCZ_PRS,³⁸ 4% for BD_PRS,³⁹ 3% for MD_PRS,⁴⁰ and 5.2% for IQ_PRS.⁴¹

PRSs Comparisons Between Groups. For PRS comparisons, we restricted our analysis to subjects with genetically European ancestry because the other groups

were underpowered for such analyses. As a secondary analysis, we comprised all the multi-ancestry samples. We ran a multinomial logistic regression to estimate the effect of SCZ, BD, MD, and IQ PRSs in predicting group membership, including 10 PCs, age, sex, and the country as covariates and using controls as the reference category. The Nagelkerke R^2 estimated the difference between models including PRS and models including covariates only. We then repeated this analysis by excluding controls and using the high-cognitive-functioning cluster as the reference category. Finally, a multivariate general linear model tested if SCZ, BD, MD, and IQ PRSs differed by the subgroups. PRSs for the graphs were adjusted by age, sex, country, and the 10 PCs. Of note, this analysis allows us to take into account all the different PRSs included. Finally, according to our secondary aim, a higher Cannabis Use Disorder PRS (CUD_PRS)⁴² could increase the probability of being a daily cannabis user.⁴³ A posthoc exploratory analysis compared clusters and controls by CUD_PRS, adjusted for the same variables. Finally, we tested Height_PRS as a control analysis.

Results

Representativeness and Characteristics of the Sample

The sample comprised 802 patients with FEP (70.9% of the original sample, $N = 1,130$) and 1,263 population controls (84.4% of the original sample, $N = 1,497$), assessed with WAIS and PAS. The patients included in this analysis were younger compared to those not assessed patients [$t(1,128) = 3.6$, $P < .001$]. There were no meaningful differences between the controls and those not included ([Supplementary Table 1](#)). However, patients differed from controls, as detailed in [Supplementary Table 2](#). **Cluster Results**

Our optimal cluster analysis identified a four-cluster solution (obtained 36.5% of times; intermediate silhouette = 0.3; BIC = 2268.5; the ratio of distance measures = 1.591) ([Supplementary Figure 1](#)). We termed the four clusters: high-cognitive-functioning, low-cognitive-functioning, intermediate, and deteriorating. Agreement in the assignment of subjects to the clusters was moderate (Fleiss' kappa = 0.419, 95% CI: 0.418, 0.420, $P < .005$), ranging from good in the deteriorating cluster, moderate in the high- and low-functioning clusters, and poor in the intermediate cluster ([Supplementary Figure 2](#)). Self-ascribed ethnicity had negligible descriptive power (predictor importance = 0.04). The high-cognitive-functioning cluster comprised 205 (25.5%) patients, who had the highest IQ ($M = 106.1$, 95% CI: 104.3, 107.9). They showed high PAF_{<12}, slightly deteriorating over time ($M_{\text{diff}} = -0.10$, 95% CI: -0.20, -0.01), but low PSF_{<12}, somewhat improving in early adolescence ($M_{\text{diff}} = 0.24$, 95% CI: 0.14, 0.33). Compared with controls, they had similar IQ ($M_{\text{diff}} = 2.7$, 95% CI: -0.4, 6),

and a higher PAF_{<12} (M_{diff} = 0.28, 95% CI: 0.16, 0.4) and PAF₁₂₋₁₆ (M_{diff} = 0.21, 95% CI: 0.02, 0.39).

Two hundred twenty-three patients (27.8%) clustered in the low-cognitive-functioning group, with the lowest IQ (M = 73.9, 95% CI: 72.2, 75.7), and PAF_{<12} slightly improving in early adolescence (M_{diff} = 0.51, 95% CI: 0.41, 0.6), but at the lowest level among groups. Their PSF_{<12} was better than PAF_{<12} and near normal but marginally deteriorating over time (M_{diff} = -0.13, 95% CI: -0.28, 0.04).

Two hundred twenty-four patients (27.9%) showed an intermediate profile, compared with the previous two groups, in terms of IQ (M = 80.8, 95% CI: 79.1, 82.5) and PAF_{<12}. They had low PSF_{<12}; however, they improved in both PAS (M_{diff} = 0.13, 95% CI: -0.03, 0.22) and PSF (M_{diff} = 0.24, 95% CI: 0.15, 0.33) in early adolescence. They differed from the high-cognitive-functioning group only in having lower IQ and PAF at both ages.

Finally, 150 patients (18.7%) were assigned to the deteriorating group, having an intermediate IQ (M = 80.6, 95% CI: 78.5, 82.7). However, in contrast to the intermediate group, started with good PAF_{<12} (M_{diff} = -0.001, 95% CI: 0.13, -0.13) and a PSF_{<12} (M_{diff} = -0.013, 95% CI: 0.13, -0.19) comparable to controls; then, they deteriorated in early adolescence in both PSF (M_{diff} = -0.74, 95% CI: -0.85, -0.62) and PAF (M_{diff} = -0.96, 95% CI: -0.08, -0.85).

Repeated measures showed that all the patient groups had more significant change between childhood and early adolescence than in the general population controls; this change was more marked in the deteriorating group than in the other three clusters (table 1; figure 1; Supplementary Tables 3 and 4).

Subgroup Comparisons

The distribution by sex was similar among clusters [$\chi^2(3) = 6.6, P = .084$]. The low-cognitive-functioning

Table 1. Comparisons in PSF, PAF, and IQ Scores

Between-groups pairwise comparisons		Within-patient statistics ^a	
Dependent variable	Contrasts	Statistic F(3,789)	P-value
IQ	C = H > I = D > L	221.7	2.5 ⁻¹⁰
PAF _{<12}	H > C = D > I > L	316.2	8.81 ⁻¹²
PAF ₁₂₋₁₆	C > H > I > D > L	121.1	1.5 ⁻⁶⁴
PSF _{<12}	C = D > L > H = I	18.7	8.91 ⁻¹³
PSF ₁₂₋₁₆	C > H = I = D = L	3.2	.021

C = CONTROLS; H = HIGH-COGNITIVE-FUNCTIONING; I = INTERMEDIATE; D = DETERIORATING; L = LOW-COGNITIVE-FUNCTIONING

Accounting for sex, age, country, and self-ascribed ethnicity.

^aStatistics are relative to within-patient comparisons by assuming the high-cognitive-functioning group as the baseline category.

and deteriorating groups were the youngest [$F(3) = 6.5, P < .001$], and the deteriorating group had the lowest percentage of European ancestry subjects [70%, ($\chi^2(3) = 19.6, P > 0.001$)], among those genotyped. Detailed between-group comparisons are in [Supplementary Table 5](#). Patients in the low-cognitive-functioning (34.8%) and the deteriorating (36.9%) group had the highest percentage of daily cannabis smokers. Among patients, daily users of high-potency cannabis were more likely to belong to the deteriorating cluster (24.8%; log ODD = 0.422, 95% CI: 0.031, 0.834, $P = .034$) [$\chi^2(3) = 11.8, P = .008$]. The distribution of specific diagnoses [$\chi^2(39) = 38, P = .515$] was similar between different clusters ([Supplementary Table 6, Supplementary Figure 3](#)). IQ subtests showed that processing speed was impaired in all clusters as compared to controls [$F(4,2047) = 195.9, P < .001$] ([Supplementary Table 7](#)). Finally, there were between-cluster differences in the distribution of the negative [$F(3,778) = 4.1, P = .006$], manic [$F(3,778) = 2.8, P = .037$] and depressive [$F(3,778) = 6.1, P = .0003$] symptom dimensions ([Supplementary Tables 8 and 9](#)). For the PRS analyses, 623 FEP and 1,000 controls, among those included in the study were genotyped. Twelve subjects were excluded from the original EU-GEI sample due to a sex discrepancy. The PRSs we measured (SCZ_PRS, BD_PRS, MDD_PRS, and CUD_PRS) were all positively related to each other and SCZ_PRS and MDD_PRS were negatively related to IQ_PRS ([Supplementary Table 10](#)); additionally, SCZ_PRS, BD_PRS, MDD_PRS, and CUD_PRS were all negatively related with PAF_{<12} and PAF₁₂₋₁₆, and IQ, whereas a positive correlation was present between these scores and IQ_PRS ([Supplementary Table 11](#)).

After ancestry restriction to European subjects only, there were 139 high-cognitive-functioning, 142 intermediate, 77 deteriorating, and 133 low-cognitive-functioning patients compared with 836 controls. Compared to the covariates-only model, ($\Delta\chi^2 = 271, df = 68, P = 4.8546^{-26}, R^2 = 0.205$), the introduction of PRSs in the clusters-control analysis explained additional variance ($R^2 = 0.079$) mostly due to SCZ_PRS (2-log Likelihood of reduced model $\chi^2_{(2-\text{Log})}(4, 20) = 48.8, P = 6.18^{-10}$), BD_PRS [$\chi^2_{(2-\text{Log})}(4, 20) = 18.4, P = .001$], and IQ_PRS [$\chi^2_{(2-\text{Log})}(4, 20) = 10.9, P = .027$]. MD_PRS did not explain additional variance in the model [$\chi^2_{(2-\text{Log})}(4, 20) = 7.5, P = .108$].

Pairwise comparisons revealed differences between controls and subgroups of patients in specific PRSs, with no substantial within-clusters differences ([table 2](#)). SCZ_PRS was higher in all clinical groups compared with controls, with the highest SCZ_PRS in the low- and high-cognitive-functioning patient groups and the lowest in the deteriorating group [$F(4, 1,322) = 18.5, P > 0.001$]. The intermediate- and high-cognitive-functioning groups had higher BD_PRS than controls, while the low-cognitive-functioning group had the lowest BD_PRS [$F(4, 1,322) = 8.9, p > 0.001$]. The high-cognitive-functioning

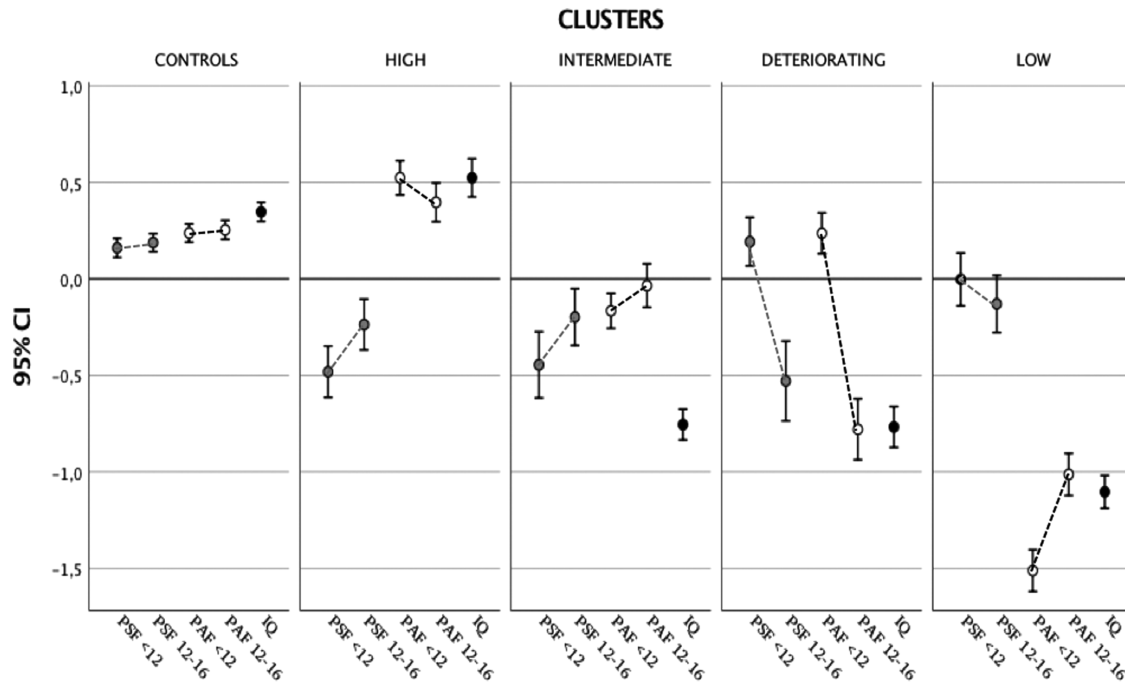


Fig. 1. Trajectories of PSF and PAF, and IQ in childhood and early adolescence by clusters of FEP and controls. Legend: Y-axis represents z-scores for PSF, PAF, and IQ in each cluster of patients with FEP and controls. HIGH = high-cognitive-functioning; LOW = low-cognitive-functioning.

Table 2. Model Summary of Group Comparisons in Terms of PRSs

	Controls	High	Intermediate	Deteriorating	Low	η^2	Contrasts	Within-patient statistics ^a	
PRS	(N = 836)	(N = 139)	(N = 142)	(N = 77)	(N = 133)			F	P
SZ	-0.185 -0.25, -0.11	0.343 0.18, 0.50	0.285 0.12, 0.44	0.184 -0.02, 0.39	0.362 0.20, 0.52	0.059	L = H = I = D > C	0.65	.583
BD	-0.113 -0.18, -0.04	0.218 0.06, 0.37	0.313 0.15, 0.46	0.192 -0.02, 0.40	0.084 -0.07, 0.24	0.028	I = H > C = D = L	1.22	.301
MD	-0.089 -0.15, -0.02	0.025 -0.13, 0.18	0.113 -0.04, 0.27	0.071 -0.14, 0.28	0.303 0.14, 0.46	0.017	L > C = I = D = H	2	.112
IQ	0.072 0.03, 0.14	0.015 -0.14, 0.17	-0.127 -0.28, 0.03	-0.07 -0.29, 0.14	-0.268 -0.43, -0.10	0.013	L < C = H = D = I	1.9	.123

Estimated Marginal Means and 95% CIs. The model is corrected by age, sex, country, and the ten ancestry PC as covariates and weighted by case/control.

High = high-cognitive-functioning; Low = low-cognitive-functioning

^aStatistics are relative to within-patient only comparisons by assuming the high-cognitive-functioning group as the baseline category. Between-patient contrasts are represented in the “contrasts” column.

and deteriorating groups had the most similar genetic predisposition to an average IQ, closer to controls. As expected, the low-cognitive-functioning group presented a worse IQ PRS [$F(4, 1,322) = 4, P = .003$] but unexpectedly higher MD_PRS than controls (table 2; figure 2; Supplementary Table 12).

PRS scores did not differ in between-cluster comparisons in FEP only, and the four PRSs explained only 2.7% of the between-clusters variance. These results were

primarily consistent with the total multi-ancestry sample, but the deteriorating sample had a higher BD_PRS than the restricted sample (Supplementary Table 14, Supplementary Figure 4). Moreover, a sensitivity analysis verified the tenure of results by running the same PRSs at the threshold of 1.0, and the results stayed consistent (Supplementary Figure 5).

Posthoc analysis on CUD_PRS [$F(4, 1,303) = 3.25, P = .011, \text{effect size} = 0.010$] revealed higher scores

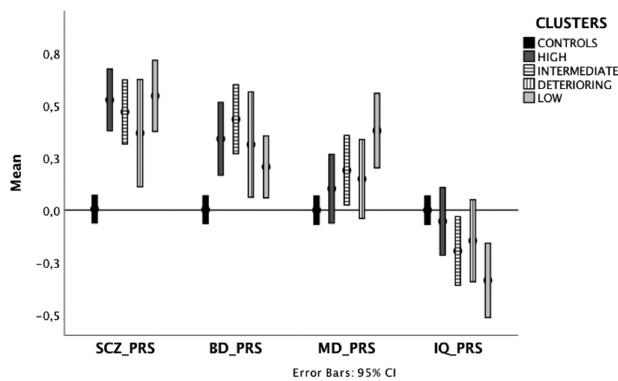


Fig. 2. Polygenic risk scores across different clusters of FEP and controls. Legend: The Y-axis indicates z-scores for PRSs. Higher standardized scores indicate higher genetic risk for the SCZ_PRS, BD_PRS, and MD_PRS. For IQ_PRS, lower standardized scores indicate a predisposition to worse IQ. All the PRSs were adjusted to account for age, sex, country, and ten PCs and standardized around controls' means and standard deviations. Each controls' PRS represents a mean of 0 and a standard deviation of 1. Error bars represent 95% CIs. HIGH = high-cognitive-functioning; LOW = low-cognitive-functioning.

in the low-cognitive-functioning group with respect to controls ($M_{diff} = -0.000113$, 95% CI: -0.000215 , $-1.20E-05$, $P = .017$) but no within-cluster differences (Supplementary Table 13, Supplementary Figure 6). Finally, the Height_PRS that we used as a control PRS was not related to group membership [$F(4, 1,274) = 2.122$, $P = .079$].

Discussion

Main Findings

Using cluster analysis of patients based on the premorbid adjustment and current IQ, we identified four independent clusters, in line with previous studies with mixed-diagnosis samples.⁵ However, unlike previous studies, we did not find segregation of those with bipolar diagnoses into the high-cognitive-functioning group and schizophrenia in the intermediate and low-cognitive-functioning groups, nor more symptoms in patients with lower IQ.^{5,44} Recruiting patients at FEP may have considerably overcome the problems of illness-related social impairment and symptom duration influencing cognitive abilities. Moreover, research on those with long-established diagnoses¹¹ could not have studied the entire patient sample included in FEP studies because a proportion of patients with psychosis recovered.⁴⁵ PRSs explained 7.9% of the total variance in comparing the four clusters to controls and 2.7% of the variance of the within-patient cluster. Interestingly, the cluster of patients showing functional deterioration did not have a higher genetic liability to schizophrenia than other groups of patients.

However, this cluster was more likely composed of heavy cannabis users. Altogether, these findings might

reflect that deterioration is a complex phenomenon involving other genetic culprits and environmental risk factors, able to reduce the subject's social and cognitive capacities.²⁴

Finally, we did not find more cognitive impairment among patients with higher SCZ_PRS.⁴⁶⁻⁴⁹ Instead, cognitive function was mainly related to variations in IQ_PRS.⁵⁰

The High Cognitive-Functioning Cluster

In line with other studies,⁴ 25% of patients fell into the high-cognitive-functioning cluster. They were well-educated, cognitively preserved, apart from processing speed,^{1,10,51-53} and outperformed other patients and controls in PAF, total IQ, and cognitive subdomains. Patients in this group had the highest between-patients IQ_PRS and very high SCZ_PRS. In contrast to their preserved cognitive function, this group of patients showed childhood impairment in PSF, slightly improving in adolescence, possibly impairing social problem-solving capacities.⁵⁴ The Low-Cognitive-Functioning Cluster

The low-cognitive-functioning group included 28% of patients, similarly to studies describing patients with constant poor-cognitive-functioning before the onset,⁴ rather than deterioration, as was reflected in their PAF.¹⁶ The inclusion of all those with reasonably reliable IQ scores (≥ 45) in our study⁵⁵ described a genuinely low-cognitive-functioning group, probably enclosing neurodevelopmentally impaired patients.^{56,57} However, they were not necessarily the most globally compromised over their lifespan. They showed better PSF in childhood than high-cognitive-functioning and intermediate patients, although slightly deteriorating in early adolescence. This group was the most different from controls in terms of polygenic liability. In addition to higher SCZ_PRS, this cluster presented lower IQ and higher MD_PRSs than controls. The first finding confirms the role of IQ as an independent risk factor for developing psychosis,^{50,58} and at an earlier age of onset.^{59,60} According to emerging evidence,⁶¹ some people with a high MD_PRS and a neurodevelopmental genetic vulnerability (ie, higher SCZ and lower IQ PRS) may develop psychosis, whereas others develop major depression at similar ages.⁶² This differentiation raises an interesting question about the role of additional risk factors.⁶³ Notably, almost 35% of patients in this group smoked cannabis daily, which at least triple the odds of developing psychosis²⁴ and contributes to lower age of onset.^{64,65} Accordingly, CUD_PRS in the low-cognitive-functioning group was the highest among clusters and differentiated them from controls.

The Intermediate Cluster

The intermediate group included 28% of our patients. Accordingly, their IQ PRS was effectively intermediate between the high and deteriorating groups from one side

and the low group from the other. Despite their lower intellectual functioning (Supplementary Table 13), intermediate patients reported the same employment rates as the high-cognitive-functioning patients. Thus, they probably had adequate functioning before onset.⁶⁶ However, this group presented SCZ_PRS and BD_PRS higher than controls, the latter being the highest among the patient groups. Unlike previous studies,^{4,15} we found a precise premorbid characterization of this group, different from the other clusters: like the high-cognitive-functioning group, these patients presented very low PSF but, like the low-cognitive-functioning cluster, they presented low PAF; both these scores improved over time, as also occurred in the control group. The lowest educational attainment, a higher percentage of non-white subjects, and differential national representativeness of this group (mainly from Spain and Brazil), compared with the high cognitive-functioning cluster (the UK and the Netherlands), could indicate that their intermediate IQ reflects the combined effect of genetic, cultural, and other environmental differences.^{21,67,68} The Deteriorating Cluster

These patients had normal childhood function (PSF and PAF) similar to controls but showed deterioration during early adolescence. The SCZ_PRS in this cluster was higher than controls, but the lowest among patients and PRSs for IQ, BD, and MD were not different from controls. In line with a recent study with non-schizophrenic older adults,⁶⁹ this finding contradicts the hypothesis of an intrinsic cognitive deterioration due to a higher schizophrenia genetic predisposition. Dickinson et al.¹¹ revealed higher SCZ_PRS and lower IQ_PRS in patients with “adolescent declining” from premorbid to current IQ. However, they excluded patients with a history of substance abuse. We observed that deteriorating patients were likelier to have smoked high-potency cannabis daily, both known risk factors for psychosis.²⁴ This pattern could be responsible for the lower age of onset of this group,^{64,65} as compared with the high-cognitive-functioning and the intermediate group, and could have contributed to deteriorating PAF and, hypothetically, current IQ as described in our previous study.²² However, their CUD_PRS was not different from controls and was lower than the low- the high-cognitive-functioning patients. This finding is in line with previous analyses suggesting that the probability of smoking high-potency cannabis, which increases tolerance and the frequency of cannabis use, is more dependent on the market availability than on any polygenic predispositions.^{23,24} Also, nearly half of patients in this cluster ascribed to a minority ethnic group, which may have represented an additional risk factor reflecting sociocultural disadvantage.^{70–72} Nevertheless, the current study cannot provide definitive causal evidence of the relationship between heavy cannabis use and a deteriorating pattern in premorbid adjustment. Cannabis was the only environmental risk factor explored because of its relationship

to cognition and premorbid adjustment.²² Additionally, environmental risk factors often correlate with other environmental and polygenic risk factors,⁷³ like cannabis abuse and minority ethnic groups, urbanization, and social isolation.^{74,75} Thus, one of the most recent approaches using environmental risk score algorithm⁷⁶ would help explore the hypothesis of higher environmental exposure in this cluster of FEP.

Nonetheless, we have to acknowledge some limitations of the study. First, there were marginal variations in the PRS profile between-patient groups. Although the GWASs used different power and predictive values,⁷⁷ we cannot directly compare the discriminative ability between different PRSs. Higher power would be necessary to find differences between groups with high-superimposable genetic characteristics.⁷⁸ PRSs should be further considered as dynamic sources of weighting effects of risk allele effects, as they vary each time more powered GWASs are released. In the interim, we may consider the lack of additional explanatory power of the MD_PRS⁴⁰ is due to its lower predictive value, which is less than half that SZ_PRS.³⁸

Additionally, differences could emerge in more sophisticated analyses based on subsets of genes or addressing the role of other components of the human genomic variation (eg, copy number variations), which could be present more in a cluster than in another.⁷⁹ The restriction of ancestry to solely white Europeans excluded a group of subjects, particularly at risk for psychosis, and limited the generalizability of our findings.⁷⁰ Furthermore, it is a cross-sectional study, and the reconstruction of the premorbid development is subject to self-report. However, this is not especially critical when using the PAS.⁸⁰ Equally, it would be interesting to look at the cognitive trajectories of our clusters at follow-up. The clusters resulting from a multicultural sample could only be partially replicated in different subsamples. However, the variability resulting from multi-site samples yields light on differences that studies using different methodologies in more homogeneous samples should not be able to catch. Finally, we only looked at cannabis use as an environmental risk factor and did not have the opportunity to control for socioeconomic status. Nonetheless, in future directions of this study, we could examine the distribution of other environmental risk factors between clusters.

This is, to our knowledge, the first study clustering patients with FEP by premorbid and cognitive-functioning, which suggests four discrete subgroups across the psychosis spectrum. First, cognitive impairment and premorbid deterioration were not reflected in a higher polygenic risk for schizophrenia. Indeed, the group of low-cognitive-functioning at FEP seemed to reflect neurodevelopmental impairment but no childhood social impairment. In contrast, the high-cognitive-functioning group showed childhood social impairment. Second, an intermediate cognitive group showed poor but improving

childhood social and academic functioning. In contrast, a deteriorating group showed a marked deterioration in social and academic function in early adolescence, from normal functioning in childhood, probably related to an environmental impact, such as daily use of high-potency cannabis.

This study argues against the view that a pronounced and generalized deterioration in premorbid adjustment is typical of most patients who develop psychosis, especially in those more strongly predisposed to schizophrenia. Our findings could also inform care and prevention strategies⁸¹ focusing on patients' resources and on preventable risk factors, such as cannabis misuse.

Supplementary Material

Supplementary material is available at *Schizophrenia Bulletin* online.

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