Genome-wide association study across five cohorts identifies five novel loci associated with idiopathic pulmonary fibrosis

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Abstract

Idiopathic pulmonary fibrosis (IPF) is a chronic lung condition with poor survival times. We previously published a genome-wide meta-analysis of IPF risk across three studies with independent replication of associated variants in two additional studies. To maximise power and to generate more accurate effect size estimates, we performed a genome-wide meta-analysis across all five studies included in the previous IPF risk GWAS. We utilised the distribution of effect sizes across the five studies to assess the replicability of the results and identified five robust novel genetic association signals implicating mTOR signalling, telomere maintenance and spindle assembly genes in IPF risk.

Introduction

Idiopathic pulmonary fibrosis (IPF) is a chronic lung disease believed to result from an aberrant response to alveolar injury leading to a build-up of scar tissue. This progressive scarring is eventually fatal with half of individuals dying within 3 to 5 years of diagnosis[1]. The cause of IPF is unknown but genetics play an important role in how susceptible an individual is to IPF[2].

Genome-wide association studies (GWAS) are an approach whereby genetic variants from across the genome are tested for their association with a disease. Genetic loci identified by GWAS can implicate genes important in disease pathogenesis and drugs which target the products encoded by these genetically-supported genes are twice as likely to be successful during development. The genetic association statistics from a GWAS are also widely used to identify causal markers of disease through Mendelian randomisation, to conduct heritability estimation and for genetic correlation analyses.

We recently published a GWAS of IPF risk[2]. The discovery GWAS consisted of three studies (named as the UK, Chicago and Colorado studies) and a replication analysis performed in two independent studies (named as the UUS [USA, UK and Spain] and Genentech studies). This analysis reported 14 genetic signals which implicated host defence, cell-cell adhesion, spindle assembly, TGF-β signalling regulation and telomere maintenance as important biological processes involved in IPF disease risk. The effect size estimates from this analysis have been widely used in other genetic analyses[3–5] and have been integrated into drug target discovery pipelines.

To maximise sample sizes for detection of new genetic associations, and to generate more precise effect size estimates, we have reanalysed the data and present a meta-analysis of genome-wide data from all 5 datasets included in our previous study. The results of this analysis implicate new genetic loci in IPF pathogenesis and provide a unique resource for other studies of IPF risk and pathogenesis.

Methods

Quality control and sample selection have been previously described[2]. In summary, datasets comprised of unrelated European-ancestry individuals from across the USA, UK and Spain, diagnosed using ATS/ERS guidelines[6,7]. Individuals in the Genentech study were sequenced using HiSeq X Ten platform (Illumina) and all other individuals were imputed from genotyping data using the HRC reference panel[8]. Genome-wide analyses were performed in each study separately using an additive logistic regression model adjusting for the first 10 genetic principal components to account for population stratification.

The five separate study-level GWAS were meta-analysed into one single GWAS, using an inverse-variance weighted fixed effect meta-analysis using METAL[9]. Variants were included in the meta-analysis if they were available in at least four studies. Genomic control was performed on the meta-analysis results using the LD score regression intercept to account for inflation not explained by polygenic effects[10]. Significant variants were defined as those with meta-analysis $p<5\times10^{-8}$ and conditional analyses were performed using GCTA-COJO to identify additional independent associated variants[11]. Independent associated variants were defined as variants remaining genome-wide significant after conditioning on the most significant variant (sentinel) in the region with consistent effect size estimates in the conditional and non-conditional analysis. Annotation of the sentinel variants was then performed using Variant Effect Predictor[12].

To assess the robustness of novel results, we tested the strength and consistency of results across studies using MAMBA (Meta-Analysis Model-Based Assessment of replicability)[13]. Variants with a posterior probability of replicability (PPR) 90% were considered robust and likely to replicate should additional independent datasets become available.

Summary statistics (i.e. effect size estimates, standard errors, p values and basic variant information) for all variants included in the genome-wide meta-analysis can be accessed at https://github.com/genomicsITER/PFgenetics.

Results

A total of 4,125 cases, 20,464 controls and 7,554,248 genetic variants were included in the analysis (Figure 1). The UUS study included one additional case (due to resolving a sample ID issue since the previous publication) and one fewer control (where the individual has since withdrawn consent from UK Biobank) than described in the previous GWAS[2].

After conditional analyses, there were 23 independent signals with $p<5\times10^{-8}$ in the genome-wide meta-analysis (Figure 2). These 23 signals included all 14 associations reported in the previous GWAS (Supplementary Table 1). Of the nine novel genetic associations (Table 1), five showed evidence of replicability (PPR 90%). The sentinel variants of these five loci included variants in introns of *KNL1*, *NPRL3*, *STMN3* and *RTEL1,* and an intergenic variant in 10q25.1. All five novel variants had consistent direction of effect across all of the individual studies and reached nominal significance ($p<0.05$) in at least 3 of the studies. Twelve of the 14 previously reported signals had PPR>90% (Supplementary Table 1).

Discussion

By increasing the number of cases in the discovery analysis by more than 50% compared with the previous IPF risk GWAS, we identified novel genetic signals associated with IPF risk and improved the precision of estimations for previously reported signals. The five novel loci had internal evidence of replicability giving us confidence that these signals are likely to be generalisable.

The signals in *RTEL1* and *OBFC1* have been reported previously but did not meet the significance criteria of the previous three-way GWAS[2]. The new MAMBA analysis suggests that the consistency of effect across studies provides high confidence that the *RTEL1* signal will replicate should an independent dataset become available. This is not the case for the *OBFC1* signal where a low posterior probability of replication suggests that there may be heterogeneity in effect across the contributing studies.

The novel signals require further characterisation to determine the likely causal gene and underlying functional effect of the variants. However, some of the genes that are closest to these new signals have strong candidacy for involvement in IPF pathogenesis. *NPRL3* encodes a GATOR1 complex function component and acts through mTORC1 signalling to inhibit mTOR kinase activity[14]. mTOR regulates TGF-β collagen synthesis and inhibiting mTOR leads to increased deposition of scar tissue[15]. We previously reported an association implicating *DEPTOR*, another mTOR inhibiting gene. We also add to the evidence that cellular ageing plays a key role in IPF pathogenesis through associations at the telomere maintenance genes *TERT*, *TERC* and *RTEL1*. We previously reported associations in spindle assembly genes (*MAD1L1* and *KIF15*) and have identified a novel genetic association in another spindle assembly gene *KNL1* (Kinetochore Scaffold 1 also known as *CASC5*). *STMN3* (Stathmin 3) implicates another cell replication process through tubulin binding[14].

Our analysis also shows the benefits of including all samples in the genome-wide analysis. By utilising recent statistical methodological advances to test for the replicability of signals

when all available datasets are included in the discovery GWAS[13], we were able to identify five additional variants with evidence of being robustly associated with IPF risk. Additional independent replication of these signals would strengthen the evidence for their role in IPF susceptibility.

By maximising the statistical power of the analysis, we identified novel genetic associations with IPF risk. These signals may implicate biologically relevant genes that support the importance of TGF-β signalling and cell replication as important processes in disease pathogenesis

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

Funding

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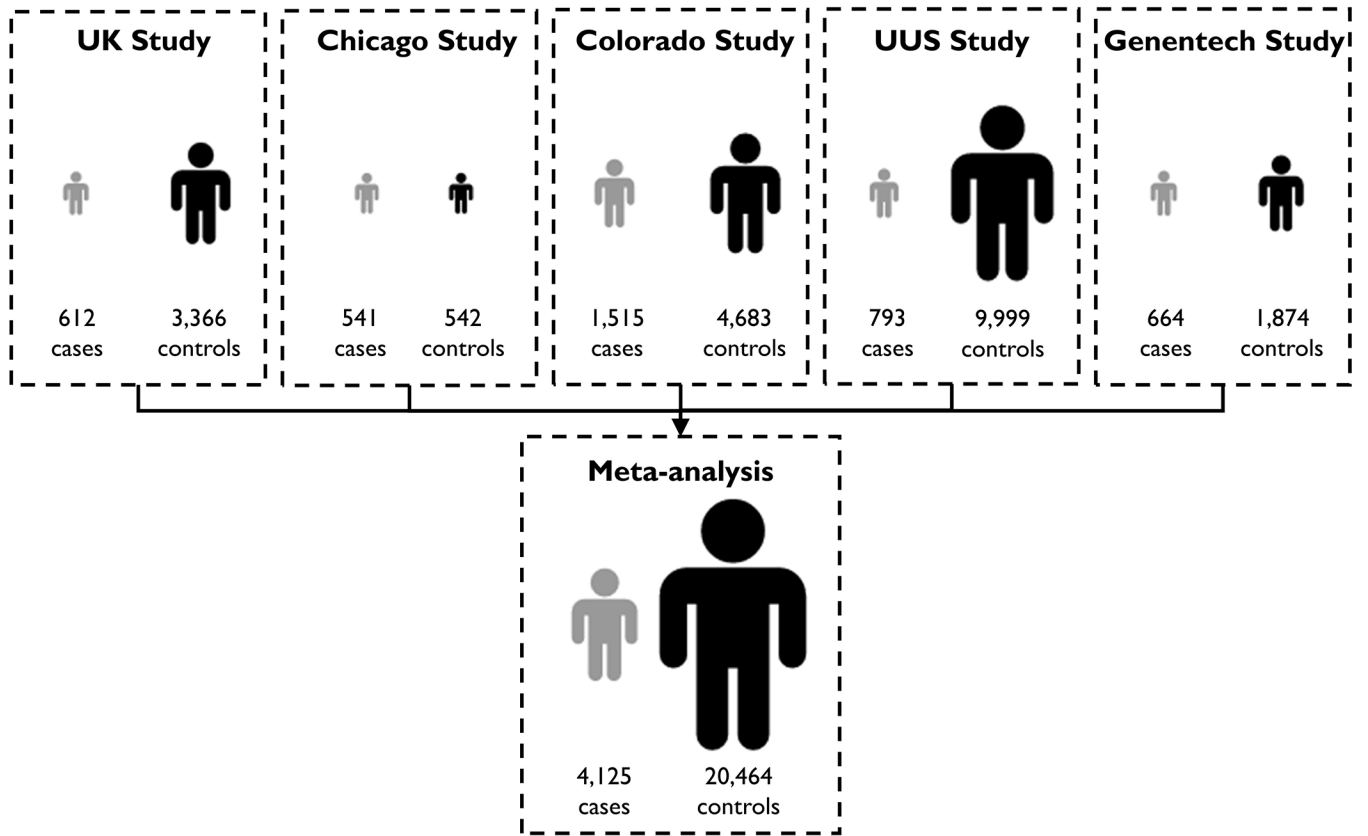
Figure 1.
Study design and sample sizes.
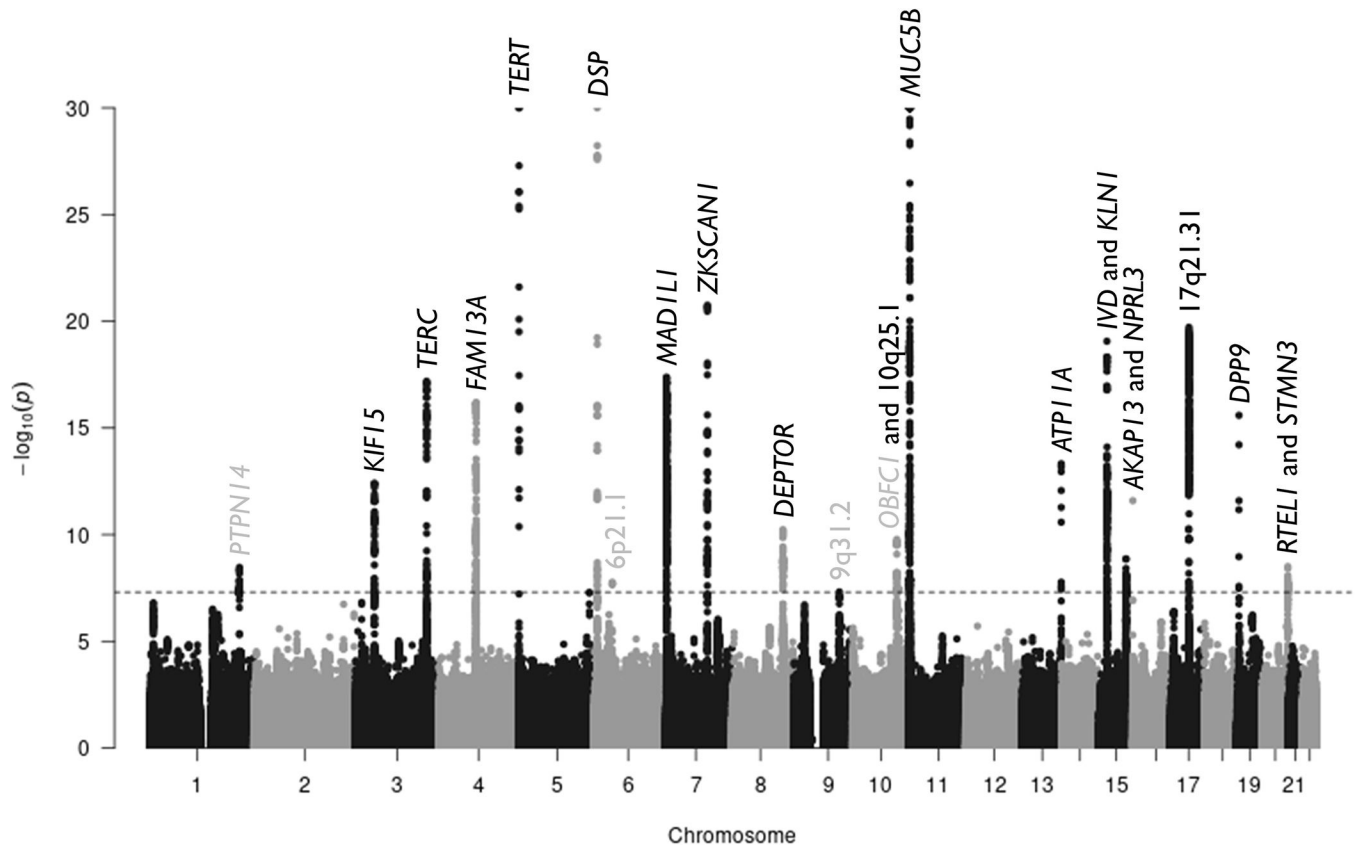
Figure 2.
Manhattan plot. Each point shows a genetic variant with chromosomal position on the x axis and the −log(p value) on the y axis. The grey dashed line shows the genome-wide significance level (p=5×10⁻⁸). Each signal is labelled with the gene implicated by that signal. Genes in grey are the novel loci that do show evidence of replicability. The plot has been truncated at p=10⁻³⁰

Table 1:

Sentinel variants of novel associations.

Chr	Position	rsid	Annotation	Ref allele	Effect allele	EAF	Direction	Study p 0.05	OR [95% CI]	p	PPR
i) Novel variants with high posterior probability of replication (PPR 90%)											
10	111229861	rs79684490	Intergenic (10q25.1)	G	A	4.6%	+ + + + +	YYNYY	1.40 [1.24, 1.57]	3.52×10^{-8}	**94.0%**
15	40931708	rs12912339 [a]	Intron of *KNL1*	G	A	15.9%	+ + + + +	YYNYY	1.30 [1.21, 1.39]	7.41×10^{-13}	**96.5%**
16	162240	rs74614704	Intron of *NPRL3*	G	A	5.6%	+ + + + +	YNNYY	1.49 [1.33, 1.67]	2.57×10^{-12}	**99.4%**
20	62284170	rs112087793 [b]	Intron of *STMN3*	T	C	91.5%	+ + + + +	YYYYY	1.34 [1.21, 1.48]	1.09×10^{-8}	**96.8%**
20	62324391	rs41308092 [b]	Intron of *RTEL1*	G	A	2.1%	+ + + + +	YYYYN	1.75 [1.45, 2.10]	3.13×10^{-9}	**99.9%**
ii) Novel variants not reaching PPR 90% threshold											
1	214659598	rs4233306	Intron of *PTPN14*	T	C	80.2%	+ + + + +	YYNNN	1.23 [1.15, 1.32]	3.41×10^{-9}	37.4%
6	43352980	rs1214759	Intergenic (6p21.2)	A	G	67.9%	+ + + + +	NYYYN	1.18 [1.11, 1.25]	1.71×10^{-8}	21.9%
9	109480268	rs11788059	Regulatory region variant (9q31.2)	T	C	34.2%	+ + + + +	NYNYY	1.17 [1.10, 1.23]	4.85×10^{-8}	3.1%
10	105640978	rs7100920	Regulatory region of *OBFC1*	C	T	49.0%	+ + − + +	NYNYY	1.19 [1.13, 1.26]	1.67×10^{-10}	32.1%

Novel variants are defined as those not reaching significance criteria in previous analysis[2] (the *RTEL1* and *OBFC1* signals have previously shown a possible association – see discussion). Effect sizes and directions are given in terms of the allele that increases risk of IPF. Chr=Chromosome.

Position is based on genetic build 37. Annotation obtained from Variant Effect Predictor[12]. EAF=Effect allele frequency calculated across the five studies. The "Direction" column shows the direction of the beta in each of the five individual studies (+ means beta>0, − means beta<0). The "Study p 0.05" column denotes which individual studies the variant reached nominal significance in (Y means p 0.05, N means p>0.05). Both the direction and study p<0.05 are given in the order UK, Colorado, Chicago, UUS and then Genentech. OR=Odds ratio. CI=Confidence interval.

PPR=posterior probability of replicability calculated using MAMBA[13].

[a]The signal at *KNL1* is independent of the previously reported nearby signal in the *IVD* gene.

[b]The *RTEL1* and *STMN3* signals are independent of each other.