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Sixteen new lung function signals identified through 1000 Genomes Project reference panel imputation

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Lung function measures are used in the diagnosis of chronic obstructive pulmonary disease. In 38,199 European ancestry individuals, we studied genome-wide association of forced expiratory volume in 1 s (FEV₁), forced vital capacity (FVC) and FEV₁/FVC with 1000 Genomes Project (phase 1)-imputed genotypes and followed up top associations in 54,550 Europeans. We identify 14 novel loci ($P < 5 \times 10^{-8}$) in or near *ENSA*, *RNU5F-1*, *KCNS3*, *AK097794*, *ASTN2*, *LHX3*, *CCDC91*, *TBX3*, *TRIP11*, *RIN3*, *TEKT5*, *LTBP4*, *MN1* and *AP1S2*, and two novel signals at known loci *NPNT* and *GPR126*, providing a basis for new understanding of the genetic determinants of these traits and pulmonary diseases in which they are altered.

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Lung function, as measured by spirometry, predicts morbidity and mortality^{1,2}. Altered lung function is a key criterion for the diagnosis of chronic obstructive pulmonary disease (COPD), a leading cause of death worldwide^{3,4}. The ratio of forced expiratory volume in 1 s (FEV₁) over forced vital capacity (FVC) defines patients with airflow obstruction, while FEV₁ is used to assess the severity of the obstruction. Reduced FVC values are seen in restrictive lung diseases such as pulmonary fibrosis⁵. While environmental risk factors, such as tobacco smoking or air pollution, play a significant role in determining lung function^{6,7}, genetic factors are also important contributors, with estimates of heritability ranging between 39 and 54% (refs 8,9).

Genome-wide association studies (GWAS) of around 2.5 million common (minor allele frequency (MAF) > 5%) single-nucleotide polymorphisms (SNPs) in Europeans have identified 32 loci associated with lung function at genome-wide significance level ($P < 5 \times 10^{-8}$)¹⁰⁻¹⁴. However, as for other complex traits^{15,16}, these loci only explain a limited proportion of the heritability^{11,13}. Among explanations for the ‘missing heritability’ are a large number of, as yet, undetected common variants with modest effect sizes, in addition to low-frequency (1% < MAF ≤ 5%) and rare (MAF ≤ 1%) variants with larger effect sizes^{16,17}. Of particular relevance to low-frequency variants, phase 1 of the 1000 Genomes Project¹⁸ sequenced 1,092 individuals from 14 populations, providing an imputation reference panel of ~38 million SNPs and 1.4 million indels, including autosomal and X chromosome variants.

The aim of the current study, undertaken within the SpiroMeta consortium, was to improve coverage of low-frequency variants and detect novel loci associated with lung function by undertaking imputation of GWAS data to the 1000 Genomes Project¹⁸ Phase-1 reference panel in 38,199 individuals of European ancestry. We meta-analysed GWAS results across 17 studies and followed up the most significant associations with *in silico* data in up to 54,550 Europeans. We identify 14 new loci associated with lung function at genome-wide significance level, and novel distinct signals at two previously reported loci. These include two low-frequency variant association signals, which seem to be explained by non-synonymous SNPs. The results of these analyses implicate both previously considered and novel mechanisms influencing lung function.

Results

We undertook a meta-analysis of 17 GWAS imputed using the 1000 Genomes Project¹⁸ Phase-1 reference panel in a study of 38,199 individuals of European ancestry in stage 1 (Fig. 1), of which 19,532 were individuals not included in the discovery stage of previous meta-analyses of lung GWAS¹⁰⁻¹³. Characteristics of cohort participants, genotyping and imputation are shown in Supplementary Table 1. Each study adjusted FEV₁, FEV₁/FVC and FVC, for age, age², sex, height and principal components for population structure, separately for never and ever smokers. Fourteen studies additionally undertook analyses for X chromosome variants (33,009 individuals, Supplementary Fig. 1 and Methods). Inverse normally transformed residuals were then used for association testing within each smoking stratum, assuming an additive genetic effect. Within each study, we combined smoking strata association summary statistics using inverse variance-weighted fixed-effects meta-analysis, and applied genomic control¹⁹ to account for residual population structure not accounted for by principal components. We subsequently combined study-specific estimates across studies using inverse variance weighting, and applied genomic control¹⁹ after fixed-effects meta-analysis. The genomic inflation factor across autosomal variants was 1.03 for each of the three traits, and

STAGE 1 (genome-wide association studies)
n=38,199

B58C (n=5,934)
BHS1&2 (n=4,355)
CROATIA-Korcula (n=826)
CROATIA-Split (n=493)
CROATIA-Vis (n=925)
EPIC population based (n=2,339)
GS:SFHS (n=8,093)
H2000 (n=821)
KORA F4 (n=1,474)
KORA S3 (n=1,147)
LBC1936 (n=991)
NFBC1966 (n=4,563)
NSPHS (n=871)
ORCADES (n=1,802)
SAPALDIA (n=1,378)
SHIP (n=1,768)
YFS (n=419)

STAGE 2 (follow-up studies)
n=54,550

ECRHS (n=1,747)
PIVUS (n=837)
TwinsUK (n=3,023)
UK BiLEVE (n=48,943)



Figure 1 | Study design for autosomal chromosome analyses. The discovery stage (stage 1) included 17 studies and 38,199 individuals. Fifty-five variants were followed up in stage 2, which comprised four studies and 54,550 individuals.

across X chromosome variants was 1.04 for FEV₁ and 1.00 for FEV₁/FVC and FVC. Quantile–quantile plots are presented in Supplementary Fig. 2a. Variants with effective sample sizes (N effective, product of sample size and imputation quality summed across studies) < 70% were filtered out, and a total of 8,694,268 variants were included in this genome-wide study.

Forty-eight SNPs and seven indels in independent autosomal chromosome regions (± 500 kb either side of sentinel variant) with stage 1 $P < 5 \times 10^{-6}$ were followed up in stage 2 using *in silico* data from four studies comprising 54,550 individuals (Fig. 1; Supplementary Table 2). One SNP on the X chromosome also met these criteria and was followed up in a subset of three studies comprising 52,359 individuals (Supplementary Fig. 1; Supplementary Table 2). Characteristics of follow-up (stage 2) cohort participants, genotyping and imputation are shown in Supplementary Table 1. Stage-2 studies adjusted the traits for age, age², sex, height and principal components to account for population structure and ever-smoking status, and also undertook association testing on the inverse normally transformed residuals assuming additive genetic effects. Stage-2 estimates were combined across studies, and then with stage-1 estimates, using inverse variance-weighted fixed-effects meta-analysis. Thirteen SNPs and three indels, each representing new signals of association, met a genome-wide significance threshold corrected for multiple testing ($P < 5 \times 10^{-8}$) after combining stage-1 and stage-2 results (Table 1; Fig. 2), of which 10 SNPs and three indels achieved independent replication meeting a Bonferroni-corrected threshold for 56 tests ($P < 8.93 \times 10^{-4}$) in stage 2 alone.

Sixteen novel association signals for FEV₁, FEV₁/FVC and FVC.

Of the 16 novel signals reaching genome-wide significance, two represent distinct new signals for FEV₁/FVC in previously reported loci^{10,12} (stage-1 P value conditioned on previously reported variant < 5×10^{-6}). Among the remaining 14, five new loci were identified for FEV₁, six new loci for FEV₁/FVC and three new loci for FVC (Table 1). The sentinel variants at the 16 loci were in or near the following genes: *MCL1-ENSA* (1q21.3), *LYPLAL1-RNU5F-1* (1q41), *KCNS3-NT5C1B* (2p24.2), *AK097794* (3q25.32), *NPNT* (4q24), *GPR126-LOC153910* (6q24.1), *ASTN2* (9q33.1), *LHX3* (9q33.1), *PTHLH-CCDC91* (12p11.22), *TBX3* (12q24.21), *TRIP11* (14q32.12), *RIN3* (14q32.12), *EMP2-TEKT5* (16p13.13),

Table 1 | Variants associated with FEV₁, FEV₁/FVC or FVC.

rs number (chr:position)	Gene (function)	eQTL gene	Coded allele/ other allele	Measure	Stage 1				Stage 2				Meta-analysis	
					Coded allele freq.	N	Beta (s.e.)	P	Coded allele freq.	N	Beta (s.e.)	P	Beta (s.e.)	P
rs6681426 (chr1:150586971)	<i>MCL1</i> (dist. = 34,757), <i>ENSA</i> (dist. = 7,628)	<i>ARNT</i>	G/A	FEV ₁	0.36	37,944	0.042 (0.008)	1.07 × 10 ⁻⁷	0.35	5,4301	0.021 (0.006)	1.13 × 10 ⁻³	0.029 (0.005)	4.35 × 10 ⁻⁹
rs201204531 (chr1:219963090)	<i>LYPLAL1</i> (dist. = 576,883), <i>RNUSF-1</i> (dist. = 83,529)	—	A/ATG	FEV ₁ /FVC	0.41	34,866	-0.038 (0.008)	1.98 × 10 ⁻⁶	0.38	53,760	-0.027 (0.006)	1.77 × 10⁻⁵	-0.031 (0.005)	2.68 × 10 ⁻¹⁰
rs61067109 (chr2:18292452)	<i>KCNK3</i> (dist. = 178,227), <i>NTSC1B</i> (dist. = 443,537)	—	G/A	FEV ₁ /FVC	0.77	37,416	-0.050 (0.009)	3.09 × 10 ⁻⁸	0.77	54,341	-0.042 (0.007)	6.56 × 10⁻⁹	-0.045 (0.006)	1.40 × 10 ⁻¹⁵
rs6441207 (chr3:158282459)	<i>AK097794</i> (ncRNA_intronic)	<i>MLF1</i>	C/T	FVC	0.59	36,173	0.036 (0.008)	4.54 × 10 ⁻⁶	0.59	53,174	0.036 (0.006)	5.85 × 10⁻⁹	0.036 (0.005)	1.27 × 10 ⁻¹³
rs6856422 (chr4:106841962)	<i>NPNT</i> (intronic)	—	G/T	FEV ₁ /FVC	0.53	31,446	-0.044 (0.008)	1.30 × 10 ⁻⁷	0.56	49,026	-0.055 (0.006)	1.15 × 10⁻¹⁷	-0.051 (0.005)	1.51 × 10 ⁻²³
rs148274477 (chr6:142838173)	<i>GPR126</i> (dist. = 70,770), <i>LOC153910</i> (dist. = 9,419)	—	C/T	FEV ₁ /FVC	0.98	30,398	-0.161 (0.027)	2.68 × 10 ⁻⁹	0.97	50,047	-0.162 (0.019)	5.63 × 10⁻¹⁸	-0.162 (0.015)	9.58 × 10 ⁻²⁶
rs34886460 (chr9:119359372)	<i>ASTN2</i> (intronic)	—	T/TA	FEV ₁ /FVC	0.53	37,567	0.041 (0.008)	6.75 × 10 ⁻⁸	0.54	53,920	0.025 (0.006)	4.07 × 10⁻⁵	0.031 (0.005)	4.72 × 10 ⁻¹¹
rs2274116 (chr9:139094805)	<i>LHX3</i> (exonic)	<i>QSOX2</i>	C/T	FVC	0.67	32,004	0.041 (0.009)	2.68 × 10 ⁻⁶	0.66	51,730	0.038 (0.007)	4.20 × 10⁻⁹	0.039 (0.005)	5.55 × 10 ⁻¹⁴
rs11383346 (chr12:28283187)	<i>PTHLH</i> (dist. = 158,271), <i>CCDC91</i> (dist. = 126,946)	<i>CCDC91</i>	A/AT	FVC	0.41	37,254	-0.043 (0.008)	3.83 × 10 ⁻⁸	0.41	53,931	-0.041 (0.006)	4.42 × 10⁻¹¹	-0.042 (0.005)	9.52 × 10 ⁻¹⁸
rs10850377 (chr12:115201436)	<i>TBX3</i> (dist. = 79,467), <i>MED13L</i> (dist. = 1,194,945)	—	G/A	FEV ₁	0.67	37,268	-0.047 (0.008)	6.68 × 10 ⁻⁹	0.65	52,722	-0.028 (0.006)	1.55 × 10⁻⁵	-0.035 (0.005)	2.50 × 10 ⁻¹²
rs7155279 (chr14:92485881)	<i>TRIP11</i> (intronic)	<i>ATXN3</i>	G/T	FEV ₁	0.64	37,691	-0.041 (0.008)	1.39 × 10 ⁻⁷	0.64	54,471	-0.022 (0.006)	4.17 × 10⁻⁴	-0.030 (0.005)	1.41 × 10 ⁻⁹
rs117068593 (chr14:93118229)	<i>RIN3</i> (exonic)	—	C/T	FEV ₁	0.82	34,496	-0.048 (0.010)	2.72 × 10 ⁻⁶	0.81	52,572	-0.027 (0.008)	5.68 × 10⁻⁴	-0.035 (0.006)	2.25 × 10 ⁻⁸
rs12149828 (chr16:10706328)	<i>EMP2</i> (dist. = 31,789), <i>TEKT5</i> (dist. = 15,033)	—	G/A	FEV ₁ /FVC	0.83	33,999	0.049 (0.010)	3.16 × 10 ⁻⁶	0.83	50,807	0.035 (0.008)	3.33 × 10⁻⁵	0.040 (0.007)	7.65 × 10 ⁻¹⁰
rs113473882 (chr19:41124155)	<i>LTBP4</i> (intronic)	—	T/C	FEV ₁ /FVC	0.98	32,207	-0.174 (0.033)	1.48 × 10 ⁻⁷	0.99	52,907	-0.138 (0.028)	1.00 × 10⁻⁶	-0.153 (0.021)	9.95 × 10 ⁻¹³
rs134041 (chr22:28056338)	<i>MIAT</i> (dist. = 983,898), <i>MNI</i> (dist. = 87,927)	—	T/C	FEV ₁	0.43	37,669	-0.045 (0.008)	4.19 × 10 ⁻⁹	0.44	52,770	-0.018 (0.006)	4.19 × 10 ⁻³	-0.028 (0.005)	3.03 × 10 ⁻⁹
rs7050036 (chrX:15964845)	<i>APIS2</i> (dist. = 91,708), <i>GRPR</i> (dist. = 176,579)	—	T/A	FEV ₁ /FVC	0.38	32,285	-0.041 (0.007)	4.37 × 10 ⁻⁹	0.39	50,530	-0.013 (0.005)	1.60 × 10 ⁻²	-0.023 (0.004)	4.14 × 10 ⁻⁸

Chr., chromosome; dist., distance; eQTL, expression quantitative trait loci; FDR, false discovery rate; FEV₁, forced expiratory volume in 1s; freq., frequency; FVC, forced vital capacity. Results are shown for each sentinel variant associated ($P < 5 \times 10^{-8}$) with FEV₁, FEV₁/FVC or FVC in a joint analysis of up to 92,744 individuals of European ancestry, ordered by chromosome and position. The allele frequency presented in the table corresponds to the coded allele (Coded allele freq.), which was chosen as the reference allele in the 1000 Genomes Project reference panel, and in most instances (but not always) is the major allele. Two-sided *P* values are given for stage 1, stage 2 and the meta-analysis of both stages. Variants reaching independent replication in stage 2 ($P = 0.05/56 = 8.93 \times 10^{-4}$) are indicated with their stage-2 *P* value in bold. The sample sizes (N) shown are the effective sample sizes. The effective sample size within each study is the product of sample size and the imputation quality metric. Beta values reflect effect-size estimates on an inverse-normal transformed scale after adjustments for age, age², sex, height and ancestry principal components, and stratified by ever-smoking status. 'eQTL gene' presents a gene expressed in the lung, whose expression is associated with the highest ranking proxy for the sentinel variant (ranking on *r*² with sentinel first, all with *r*² > 0.8, and on eQTL *P*-value second, all with $P < 5 \times 10^{-5}$) with FDR < 10%.

LTBP4 (19q13.2), *MIAT-MNI* (22q12.1) and on chromosome X, *APIS2-GRPR* (Xp22.2) (Supplementary Fig. 2b,c). To gain further insight into the associated variants, we assessed whether the novel sentinel variants, or their proxies, were associated with gene expression in lung tissues²⁰ and blood²¹ (Methods, Supplementary Methods; Supplementary Table 3a) or were in DNase hypersensitivity sites²² in relevant cell types (Methods; Supplementary Table 3b). For relevant genes, we investigated RNA-seq splice isoforms in human bronchial epithelial cells (Supplementary Methods; Supplementary Fig. 3), searched for evidence of protein expression in the respiratory system²³ (Supplementary Table 3c), assessed differential expression across the pseudoglandular and canalicular stages of fetal human lung development (Methods; Supplementary Table 3d) and assessed evidence for differences in gene expression in bronchial epithelial brush samples from COPD cases and smoking controls (Methods; Supplementary Table 3e).

The two novel signals in known loci were the strongest ($P < 5 \times 10^{-23}$) association signals after meta-analysing

stage 1 and 2. The strongest signal was for a low-frequency SNP near *GPR126* (rs148274477, MAF = 2.4%, intergenic on chromosome 6) associated with FEV₁/FVC ($P = 9.6 \times 10^{-26}$, Table 1) and in high linkage disequilibrium (LD, $r^2 = 0.85$) with a missense variant (rs17280293 (Ser123Gly), Supplementary Table 3f) in *GPR126*, but distinct from the previously reported signal for FEV₁/FVC in this region^{10,13} (stage-1 *P* value for rs148274477 conditioning on rs3817928 (ref. 10) and rs262129 (ref. 13) = 1.86×10^{-7} , unconditional stage-1 $P = 2.68 \times 10^{-9}$). *GPR126* encodes a G-protein-coupled receptor and is expressed in adult and fetal lung tissue^{24,25} (Supplementary Table 3d). Other studies have shown that *GPR126* is required for mice embryonic viability and cardiovascular development²⁶, and that *GPR126* is expressed in adult mice lung²⁷. More recently, *GPR126* has been shown to bind type-IV collagen, a major collagen in the lung, leading to cAMP signalling²⁸.

The second strongest signal ($P = 1.5 \times 10^{-23}$, Table 1) was an intronic SNP (rs6856422) in *NPNT* on chromosome 4 associated with FEV₁/FVC, distinct from the previously discovered signal for

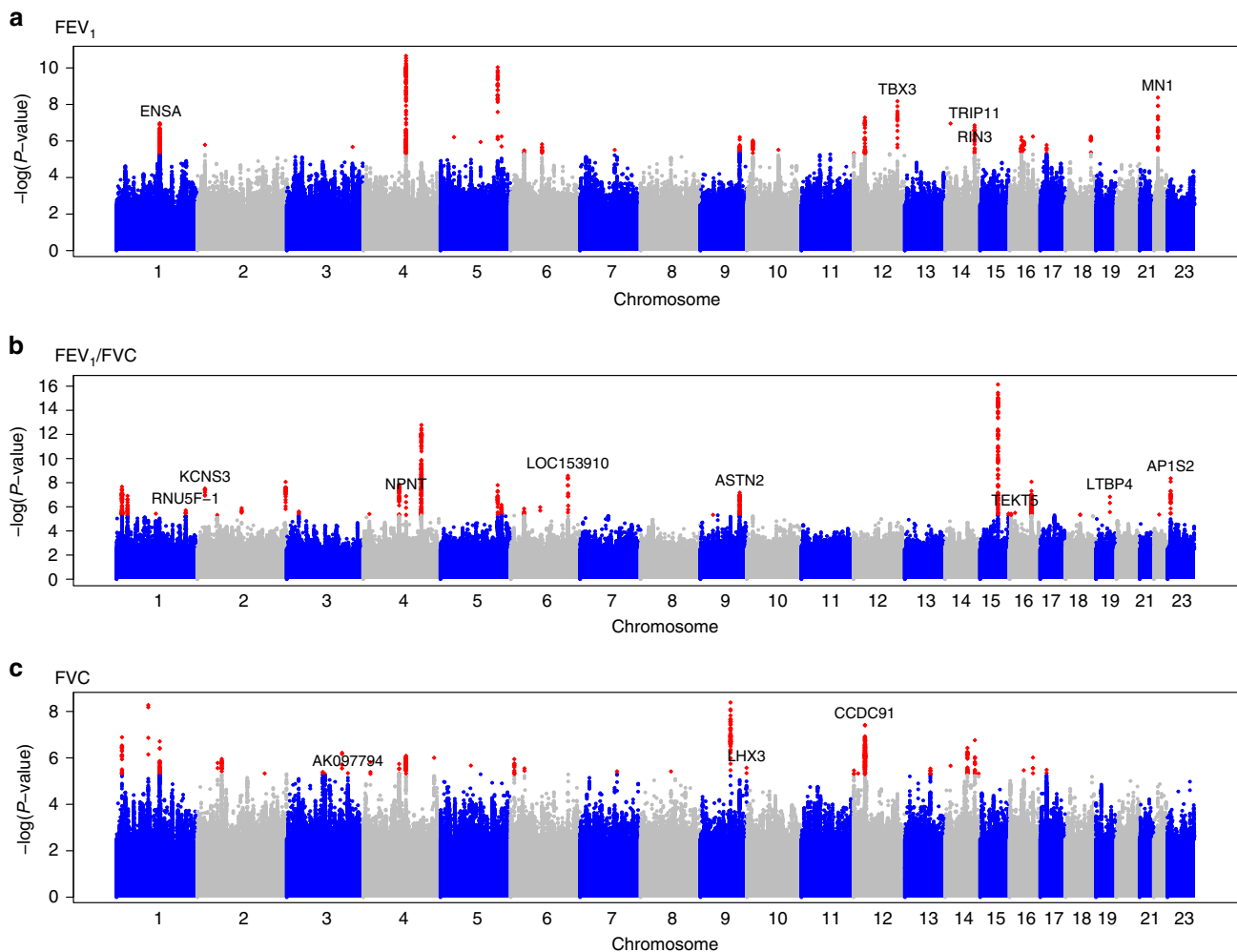


Figure 2 | Manhattan plots for association results. (a) FEV₁, (b) FEV₁/FVC and (c) FVC. Manhattan plots ordered by chromosome and position for stage-1 results. Variants with $P < 5 \times 10^{-6}$ are indicated in red. Novel signals that reached genome-wide significance after meta-analysing stage 1 and stage 2 are labelled with the nearest gene. Only variants with N effective $\geq 70\%$ are presented here.

FEV₁ in this region^{10,12,13}. The stage-1 P value for this variant conditioned on the previously reported sentinel SNPs (rs17036341 (ref. 10) and rs10516526 (refs 12,13)) and on the sentinel SNP for FEV₁ in this analysis (rs12374256, *INTS12* intron) was 4.7×10^{-6} (unconditional stage-1 $P = 1.30 \times 10^{-7}$). Proxies of the sentinel SNP were associated with expression of *INTS12* and *GSTCD* in blood (Supplementary Table 3a). *INTS12*, *GSTCD* and *NPNT* are contiguously positioned at 4q24, and are all expressed in adult and fetal lung tissues (Supplementary Table 3c,d). Our previous work characterizing *GSTCD* and *INTS12* showed that they are oppositely transcribed genes that are to some extent co-ordinately regulated, although while *GSTCD* expression in human lung tissue is ubiquitous, *INTS12* expression was predominantly in the nucleus of epithelial cells and pneumocytes²⁹.

Among the 14 novel loci, six novel loci were associated with FEV₁/FVC. One of them was a low-frequency variant (rs113473882, intronic in *LTBP4* on chromosome 19, MAF = 1.5%, Table 1) in almost complete LD ($r^2 = 0.99$) with a missense variant (rs34093919, Asp752Asn, Supplementary Table 3f) in *LTBP4*, which encodes a protein that binds transforming growth factor beta (TGF β) as it is secreted and targeted to the extracellular matrix. Mice deficient in *ltbp4* displayed defects in lung septation and elastogenesis, which may be TGF β 2 and fibulin-5 dependent³⁰, and disruption of this gene in mice led to abnormal lung development, cardiomyopathy and

colorectal cancer³¹. Variants near *LTBP4*, uncorrelated ($r^2 < 0.05$) with the sentinel SNP we report here, have been associated with COPD³² and smoking behaviour³³. A further novel FEV₁/FVC locus mapping near *AP1S2* is the first to be reported for lung function on the X chromosome; sentinel SNP (rs7050036, intergenic) proxies were associated with the expression of *AP1S2* and *ZRSR2* in lung tissue (Supplementary Table 3a). Other new loci for FEV₁/FVC were in or near *KCNS3* (2p24.2), *ASTN2* (9q33.1), *RNU5F-1* (1q41) and *TEKT5* (16p13.13).

The strongest signal for FEV₁ in a novel locus was upstream of *TBX3* on chromosome 12 (Table 1); *TBX3* is involved in the TGF β 1 signalling pathway³⁴. At a second novel locus for FEV₁ (rs7155279, *TRIP11* intron on chromosome 14, Table 1), proxies of the sentinel variant were associated with lung and blood expression of *TRIP11*. *TRIP11* encodes a protein associated with the Golgi apparatus³⁵. In the lung, rs7155279 showed strongest association with expression of *ATXN3* (Supplementary Table 3a), which encodes ataxin 3, a deubiquitinating enzyme. Expanded trinucleotide repeats in *ATXN3* cause spinocerebellar ataxia-3 (ref. 36). In blood, a proxy ($r^2 = 0.94$) for rs7155279 showed strong association ($P = 3 \times 10^{-34}$, Supplementary Table 3a) with the expression of *FBLN5*. Fibulin-5 was shown to be implicated in tissue repair in COPD³⁷ and elastogenesis and lung development³⁰. A third signal for FEV₁ was a missense variant (rs117068593, Arg279Cys, Supplementary Table 3f) in *RIN3* on

chromosome 14 (Table 1), which was ~ 632 kb from the *TRIP11* sentinel SNP (rs7155279) and independent from it ($r^2 = 8.84 \times 10^{-5}$). Although this is the first report of association of a *RIN3* variant with lung function, a correlated variant (rs754388, $r^2 = 0.99$) was recently associated with moderate to severe COPD, although the association did not replicate in an independent study³⁸. In a fourth novel region for FEV₁, on chromosome 1, a sentinel SNP, rs6681426, ~ 8 kb downstream of *ENSA* (Table 1) and a second signal ~ 700 kb apart (rs4926386, Supplementary Table 2a) were both associated with *ARNT* expression in lung (Supplementary Table 3a). *ARNT* is differentially expressed during fetal lung development (Supplementary Table 3d) and acts as a co-factor for transcriptional regulation by hypoxia-inducible factor 1 during lung development³⁹ and may regulate cytokine responses⁴⁰. SNP rs6681426 was also associated with the expression of *LASS2* (also known as *CERS2*) in lung tissue (Supplementary Table 3a); *lass2* knock-out mice develop lung inflammation and airway obstruction⁴¹. The other new locus for FEV₁ was near *MN1* (22q12.1).

All three novel loci for FVC had sentinel variants or close proxies associated with expression of a nearby gene in lung, implicating *CCDC91*, *MLF1* and *QSOX2*, located on chromosomes 12, 3 and 9, respectively. The putative function of the key genes in each of the two known and 14 novel loci for FEV₁, FEV₁/FVC and FVC are summarized in Supplementary Table 4.

Functional characterization of novel signals. The protein products of genes nearest to the sentinel variant of novel signals for lung function were expressed in bronchial epithelial cells, pneumocytes or lung macrophages (Supplementary Table 3c). Among the 16 novel signals of association with lung function, sentinel variants or close proxies were *cis* expression quantitative trait loci (eQTLs) in lung for *ARNT*, *MLF1*, *QSOX2*, *CCDC91* and *ATXN3* (Table 1; Supplementary Table 3a), and in eight loci the sentinel variant or at least one strong proxy ($r^2 > 0.8$) was in a DNase hypersensitivity site in a cell type potentially relevant to lung function (in or near *ENSA*, *RNU5F-1*, *ASTN2*, *CCDC91*, *TBX3*, *RIN3*, *TEKT5* and *MN1*, Supplementary Table 3b). The sentinel variant association was explained (conditional $P > 0.01$) by a missense variant in each of the two novel signals in which we detected a low-frequency sentinel variant (near *GPR126* and in *LTBP4*), and was explained in four of the remaining novel signals by a putatively functional variant (in or near *ENSA*, *AK097794*, *TEKT5* and *MN1*, Supplementary Table 3f and Methods). Genes in four of the novel loci showed differential expression across the pseudoglandular and canalicular stages of fetal lung development, particularly *EMP2* (Supplementary Table 3d). *MLF1* and *ATXN3* showed differences in expression levels in bronchial brushings between COPD cases and controls (Supplementary Table 3e). We detected novel splice isoforms of $> 20\%$ abundance for *GFM1*, *TRIM32*, *LTBP4* and *MN1* in human bronchial epithelial cells (Supplementary Fig. 3; Supplementary Methods).

Association in children. To assess whether the 16 new sentinel variants associated with lung function in adults may act through an effect on lung development, we assessed their association in the ALSPAC study⁴² that includes 5,062 children (Supplementary Table 5a). Eleven of the 16 sentinel variants showed consistent directions of effect in adults and children. The association with FVC of variant rs6441207 on chromosome 3 in the noncoding RNA *AK097794* exceeded a Bonferroni-corrected threshold for 16 tests (Supplementary Table 5a).

Association with smoking and gene by smoking interaction. The 16 new variants had consistent effect sizes in never smokers

and ever smokers, and no gene–smoking interaction ($P > 0.05$) in stage 1 (Supplementary Table 5b). We found no evidence that any of these signals were driven by smoking behaviour. Only the two-base-pair insertion on chromosome 1 (rs201204531) revealed an association ($P = 1.5 \times 10^{-3}$) with smoking behaviour (heavy-versus never-smoking status) that met a Bonferroni-corrected threshold for 16 tests (Supplementary Table 5c). However, this variant also showed an association with FEV₁/FVC in never smokers, and the allele associated with higher likelihood of being a smoker was associated with increased FEV₁/FVC (Supplementary Table 5b,c).

Associations with other traits. We queried the GWAS catalog⁴³ for variants in 2-Mb regions centred on the sentinel variant for the 16 loci (Supplementary Table 5d). Five loci contained variants associated with height^{44–46} (Supplementary Table 5d). In the *GPR126* and *LHX3* loci, the previously reported height variants were not correlated ($r^2 < 0.2$) with the lung function variants reported here. In the *AK097794*, *CCDC91* and *TRIP11* loci, the variants associated with height were correlated ($r^2 > 0.3$) with the lung function sentinel variants, but the alleles associated with reduced height were associated with increased FEV₁ or FVC. Associations with other traits have been reported for variants in LD ($r^2 > 0.3$) with sentinel variants in regions of *RIN3* (Paget's disease⁴⁷ and bone mineral density⁴⁸), *ENSA* (body fat mass⁴⁹ and melanoma⁵⁰) and *LHX3* (thyroid hormone levels⁵¹). None of the novel signals relate to known asthma loci, and the association findings were consistent after removing individuals with asthma (Supplementary Fig. 4).

Genetic architecture of lung function traits. The proportion of the additive polygenic variance explained by the 49 signals discovered to date (Supplementary Table 6), including new and previously reported signals^{10–14} is 4.0% for FEV₁, 5.4% for FEV₁/FVC and 3.20% for FVC (Supplementary Table 7). These estimates are likely upper bounds on the proportion of the variance explained due to the winner's curse bias. Across the 49 signals, we observed larger effect sizes for associations with lower-frequency variants (Fig. 3), supporting the hypothesis that lower-frequency variants will contribute to explaining the missing heritability¹⁶.

We examined the increase in coverage of low-frequency and common variants by the 1000 Genomes Project reference panel, compared with the HapMap imputation reference panel, at both the novel and previously reported loci (Supplementary Fig. 5a). The two association signals where the 1000 Genomes sentinel variants had low MAF ($< 5\%$), were not present when restricting the results only to variants that could be imputed using the HapMap imputation panel (rs113473882 and rs148274477 in Supplementary Fig. 5a).

For each of the 32 previously discovered regions^{10–14}, we identified the most strongly associated variant present on the 1000 Genomes Project¹⁸ reference panel and the most strongly associated variant present on the HapMap reference panel using stage-1 results, and compared the stage-1 MAFs between these two groups of variants. The 1000 Genomes sentinel variants in or near *GPR126* (rs148274477), *TGFB2* (rs147187942) and *MMP15* (rs150232756) had MAFs that were more than twofold lower than the HapMap sentinel variant MAFs (Supplementary Fig. 5b) and were statistically independent ($r^2 \leq 0.06$) from the previously discovered HapMap-imputed sentinel variants¹³. The *GPR126* 1000 Genomes-imputed sentinel was described above as one of the 16 new signals. We tested the association of the 1000 Genomes-imputed sentinel variants near *TGFB2* and *MMP15* in UK BiLEVE (Supplementary Table 8), and found supportive

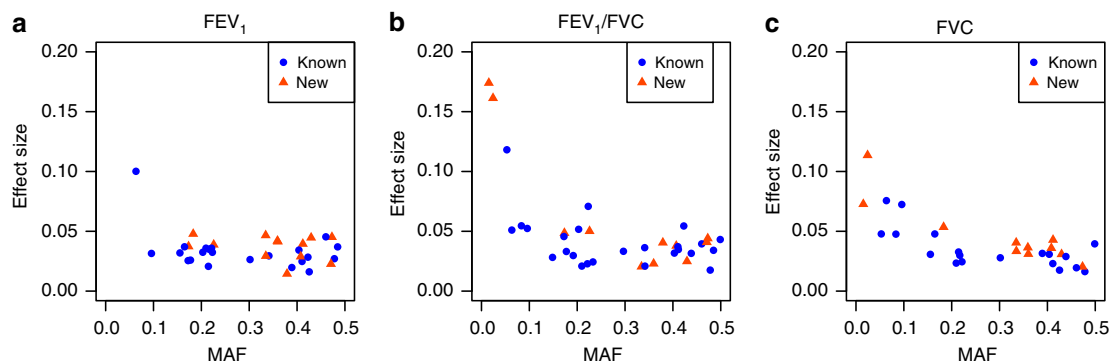


Figure 3 | Minor allele frequency against effect-size plots (a) FEV_1 , (b) FEV_1/FVC and (c) FVC. MAF is plotted against stage-1 effect sizes for variants within the 33 known^{10–14} and the 16 new signals, which had stage-1 $P < 0.05$ for association with FEV_1 , FEV_1/FVC and FVC separately. Known signals are represented with blue circles and new signals are represented with orange triangles.

evidence of association for the signal near *TGFB2* (rs147187942, MAF = 9%, $P = 5.7 \times 10^{-3}$).

Pathway analyses. We undertook a pathway analysis using MAGENTA v2 (ref. 52) and stage-1 genome-wide results for FEV_1 , FEV_1/FVC and FVC (Supplementary Methods). For FVC, the platelet-derived growth factor signalling, and the chromatin-packaging and -remodelling pathways were significant ($P = 2 \times 10^{-4}$, false discovery rate (FDR) < 0.3% and $P = 1.82 \times 10^{-4}$, FDR < 4%, respectively) (Supplementary Table 9).

Discussion

In this study, we aimed to improve coverage of low-frequency variants and detect novel loci associated with lung function, by undertaking imputation of GWAS data in 17 studies and 38,199 individuals to the 1000 Genomes Project¹⁸ reference panel, and by following up the most significant signals in an additional 54,550 individuals. Overall, 16 new association signals attained a genome-wide significance threshold corrected for multiple testing ($P < 5 \times 10^{-8}$) after meta-analysing stage 1 and stage 2, including 15 autosomal and one X chromosome signal. While two of the new findings relate to novel signals for FEV_1/FVC in previously reported regions^{10,12}, five new loci were identified for FEV_1 , six new loci for FEV_1/FVC and three new loci for FVC. Including the 16 signals discovered in these analyses, the number of lung function signals discovered to date is 49 (refs 10–14), and they jointly explain a modest proportion of the additive polygenic variance (4.0% for FEV_1 , 5.4% for FEV_1/FVC and 3.2% for FVC).

Some of the 49 distinct lung function signals^{10–14} seem to cluster close to each other. If we define regions as 500 kb either side of the sentinel variants, there are three regions that each include two distinct signals (in or near *INTS12-GSTCD-NPNT*, *GPR126* and *PTCH1* (refs 10,12)), so that the 49 signals would map to 46 loci. If we use a wider definition of region (1,000 kb either side of the sentinel), there are four regions that each include two distinct signals (in or near *INTS12-GSTCD-NPNT*, *GPR126*, *PTCH1* (refs 10,12) and *TRIP11-RIN3*). In addition, the human leukocyte antigen region on chromosome 6 includes three distinct signals (in or near *ZKSCAN3-NCR3-AGER*^{10,12,13}) within 3.8 Mb. Furthermore, we have shown evidence of an additional signal in the *TGFB2* region, and the new lung function signal in *LTBP4* lies 179 kb away from a known COPD signal³². These findings are consistent with reports from very large studies of height and lipids^{53,54}, which report multiple signals in associated regions, and highlight the importance of taking into account LD between variants to improve our understanding of

known regions. Multiple signals within known regions are likely to explain some of the hidden heritability of these traits.

To identify pathways relevant to lung function, we undertook additional analyses using MAGENTA, which have implicated pathways for platelet-derived growth factor signalling and chromatin-packaging and -remodelling. Independent analyses undertaken in a concurrent study by the UK BiLEVE consortium, which focused on the extremes of the lung function distribution⁵⁵, highlight the histone subset of the chromatin-packaging and -remodelling pathway. The *TGFβ* signalling pathway has now been implicated by three independent loci: an FEV_1/FVC signal explained by a missense variant in *LTBP4*, which encodes a protein that binds *TGFβ*; an FEV_1 signal upstream of *TBX3*, which is involved in the *TGFβ*1 signalling pathway³⁴, and a previously reported signal downstream of *TGFB2* (ref. 17). In addition, a pathway involving fibulin-5 has been implicated by two of the novel loci (*LTBP4* and *TRIP11*). The identification, through different approaches, of pathways which appear to be involved in determining lung function should help focus future functional studies.

Pathways affecting lung function also have the potential to affect COPD risk, since lung function measures are used to diagnose the disease. Currently, 13 signals (in or near *TGFB2*, *TNS1*, *RARB*, *FAM13A*, *GSTCD*, *HHIP*, *HTRA4*, *ADAM19*, *AGER*, *LOC153910*, *C10orf11*, *RIN3* and *THSD4*) out of the 49 lung function signals discovered to date^{10–14} have also shown association with some definition of COPD^{38,56–60}. This illustrates that the study of lung function measures is a powerful approach to bring insights into the genetics of COPD.

In agreement with previous findings for other lung function loci^{12,13}, none of the 16 new associations seem to be driven by either smoking behaviour or by a gene–smoking interaction. One variant showed association with smoking behaviour that met a Bonferroni correction for 16 tests in UK BiLEVE. This variant also had an effect in never smokers in stage 1, and the allele associated with increased lung function was also associated with increased risk of smoking, which does not suggest an association with lung function mediated by smoking behaviour. Variants in five out of the 16 loci associated with lung function in this study have also shown associations with height^{44–46}. However, the variants associated with height were either independent of those associated with lung function, or if they were correlated, the alleles associated with increased height, were associated with decreased FEV_1 or FVC. If the association with lung function was driven by an effect on height, we would expect consistent direction of effect between these two traits. Therefore, the associations identified for lung function in these regions are not likely to be driven by associations with height.

This study had a large follow-up stage, which included 54,550 individuals, of which 48,943 were contributed by the UK BiLEVE

study. UK BiLEVE is a particularly powerful study since it has sampled UK Biobank individuals from the extremes of the lung function distribution, and it has spirometry performed in a uniform way across individuals. Had these data been available when we undertook the discovery stage of this study, their addition would have greatly improved the discovery power. Nevertheless, incorporating these data into the follow-up stage improved power to provide replication and deal with potential winners' curse bias. Another strength of the current study design was the increased coverage of common and low-frequency variants obtained through the imputation to 1000 Genomes Project¹⁸ reference panel. This enabled us to detect two low-allele-frequency variants (with MAF of 1.5 and 2.4% and stage-1 effect sizes of 0.17 and 0.16 s.d. units, respectively) that have an effect on lung function. No associations with lower allele frequency variants have been detected in this study, despite having power > 80% in discovery to detect associations ($P < 5 \times 10^{-6}$) for variants with MAF of 0.5 and 1%, and effect sizes above 0.3 and 0.2 s.d. units, respectively. The poorer imputation quality for low-allele-frequency variants coupled with the strict criteria we used to select variants for follow-up (N effective $\geq 70\%$) have probably affected our ability to detect rare variants. For instance, a variant representing an additional signal for FEV₁/FVC in the *GSTCD-INTS12-NPNT* region, reported by the UK BiLEVE study, where it was directly genotyped⁵⁵, would have been detected in this analysis had we used a more lenient threshold (N effective > 60%). Imputation quality for rare variants will improve as larger imputation reference panels become available.

In summary, 16 new association signals for lung function have been identified in this study, including two signals explained by non-synonymous low-frequency variants. These findings highlight new loci not previously connected with lung function or COPD, and bring new insights into previously detected loci. This study also highlights the added value of imputing to new reference panels as they become available. Understanding the molecular pathways that connect the newly identified loci with lung function and COPD risk has the potential to point to new targets for therapeutic intervention.

Methods

Study design. The study consisted of two stages. Stage 1 was a meta-analysis of 17 GWAS in a total of 38,199 individuals of European ancestry. Supplementary Table 1 gives the details of these studies. Fifty-six variants selected according to the results in stage 1 were followed up in stage 2 in 54,550 European individuals.

Stage-1 samples. Stage 1 comprised 17 studies: B58C (T1DGC and WTCCC), BHS1 and -2, EPIC (obese cases and population-based studies), the EUROSPAN studies (CROATIA-Korcula, ORCADES, CROATIA-Split and CROATIA-Vis), GS:SFHS, Health 2000, KORA F4, KORA S3, LBC1936, NFBC1966, NSPHS, SAPALDIA, SHIP and YFS (see Supplementary Table 1a for the definitions of all abbreviations). All participants provided written informed consent and studies were approved by local Research Ethics Committees and/or Institutional Review Boards. Measurements of spirometry for each study are described in the Supplementary Note. The genotyping platforms and quality-control criteria implemented by each study are described in Supplementary Table 1b.

Imputation. Imputation to the all ancestries 1000 Genomes Project¹⁸ Phase-1 reference panel released in March 2012 was undertaken using MACH⁶¹ and minimac⁶² or IMPUTE2 (ref. 63) with pre-imputation filters and parameters as shown in Supplementary Table 1b. Specific software guidelines were used to impute the non-pseudoautosomal part of the X chromosome. The pseudoautosomal part of the X chromosome was not included in these analyses. Variants were excluded if the imputation information, assessed using r2.hat (MACH and minimac) or .info (IMPUTE2), was < 0.3.

Data transformation and association testing in stage 1. Linear regression of age, age², sex, height and principal components for population structure was undertaken on FEV₁, FEV₁/FVC and FVC separately for ever smokers and never smokers. The residuals were transformed to ranks and then transformed to normally distributed z-scores. These transformed residuals were then used as the

phenotype for association testing under an additive genetic model, separately for ever smokers and never smokers. For X chromosome analyses, residuals for males and females were analysed separately and dosages for males were coded 0 for 0 copies of the coded allele and 2 for 1 copy of the coded allele. The software used was specified in Supplementary Table 1b. Studies with related individuals analysed ever smokers and never smokers together adjusting the regression for ever-smoking status and used appropriate tests for association in related individuals, as described in the Supplementary Note.

Meta-analysis of stage-1 data. Quality-control checks on the stage-1 data were undertaken using GWAToolbox⁶⁴ and R version 3.0.2 (see URLs). All meta-analysis steps were undertaken using inverse variance-weighted fixed-effects meta-analysis. Effect estimates were flipped across studies so that the coded allele was the reference allele in the 1000 Genomes Project¹⁸ reference panel. For each study with unrelated individuals, autosomal chromosomes results were meta-analysed between ever smokers and never smokers. After that, all study-specific standard errors were corrected using genomic control¹⁹. Study-specific genomic inflation factor estimates are shown in Supplementary Table 1a. Finally, effect-size estimates and s.e. were combined across studies, and genomic control¹⁹ was applied again at the meta-analysis level. For the X chromosome, studies of unrelated individuals meta-analysed smoking strata estimates within sex strata and then meta-analysed pooled sex strata estimates. After that, genomic control¹⁹ was applied to each study and results were meta-analysed across studies. Genomic control¹⁹ was applied again after the meta-analysis. To describe the effect of imperfect imputation on power, for each variant we report the effective sample size (N effective), which is the sum of the study-specific products of the sample size and the imputation quality metric. Meta-analysis statistics and figures were produced using R version 3.0.2 (see URLs).

Selection of variants for stage 2. Variants with N effective < 70% were filtered out before selecting variants for follow-up (8,916,621 variants remained after filtering). Independent regions (± 500 kb from the sentinel variant) were selected for FEV₁, FEV₁/FVC and FVC if the sentinel SNP or indel had $P < 5 \times 10^{-6}$. If the same variant was selected for different traits, it was followed up for all the traits. If two different variants were selected for different traits within the same region, or if any of the regions selected had already been identified in previous GWAS¹⁰⁻¹⁴ but the sentinel variant was different from that previously reported, conditional analyses were undertaken to assess whether the signals within the same regions were distinct. If previously reported sentinel SNPs for a region were strongly correlated ($r^2 > 0.9$), we only conditioned on the SNP that had shown the strongest association. If two variants were selected for different traits within the same new region, both variants were taken forward if their P -value conditioning on the other variant was $< 5 \times 10^{-6}$; if not, only the variant with the most significant P value was taken forward. Variants within known regions were only taken forward if their P value conditioned on the previously reported variant was $< 5 \times 10^{-6}$. Conditional analyses were undertaken using GCTA⁶⁵, and B58C data were used to estimate LD. In total, 56 variants (49 SNPs and seven indels) were taken forward for follow-up, two of which were distinct signals within previously reported regions^{10,12,13}. These variants are listed in Supplementary Table 2. Previously reported signals¹⁰⁻¹⁴ were not followed up.

Stage-2 samples. The 48 SNPs and seven indels on autosomal chromosomes were followed up in up to 54,550 individuals from four studies with *in silico* data: ECRHS, PIVUS, TwinsUK and UK BiLEVE (see Supplementary Table 1a for the definitions of all abbreviations). All participants provided written informed consent and studies were approved by local Research Ethics Committees and/or Institutional Review Boards. One SNP in the chromosome X was followed up in 52,359 individuals from PIVUS, TwinsUK and UK BiLEVE. Measurements of spirometry for each study are described in the Supplementary Note.

Meta-analysis of stage-2 data. All stage-2 studies undertook linear regression of age, age², sex, height, ever-smoking status and principal components for population structure, if available, on FEV₁, FEV₁/FVC and FVC, then the residuals were transformed to ranks and to normally distributed Z-scores. These transformed residuals were then used as the phenotype for association testing under an additive genetic model. For the X chromosome analyses, allele dosages for hemizygous males were coded as 2. Effect sizes were flipped to be consistent with the stage-1 estimates, using the reference allele in 1000 Genomes Project¹⁸ as the coded allele. Genomic control¹⁹ was applied for studies that undertook the analysis genome-wide. Effect estimates and s.e. were combined across the stage-2 studies using an inverse variance-weighted meta-analysis.

Combination of stage 1 and 2 and multiple testing correction. A meta-analysis of stage-1 and stage-2 results was undertaken using inverse variance-weighted meta-analysis. We take into account the multiple tests undertaken by describing an association as genome-wide significant if it has $P < 5 \times 10^{-8}$. In addition, we assessed whether any of the findings achieved independent replication in stage 2 using a threshold corrected for the number of variants followed up ($0.05/56 = 8.93 \times 10^{-4}$).

Functional characterization of novel loci. A series of analyses were undertaken to provide insights into the expression of genes within the 16 loci (defined as ± 1 Mb either side of the sentinel variant) represented here. Blood²¹ and lung tissue²⁰ eQTL analyses were undertaken for variants in these loci that were in LD ($r^2 > 0.3$) with the sentinel variant in the region. We assessed whether variants within these loci that were strongly correlated with the sentinel variants ($r^2 > 0.8$) were in DNase hypersensitivity sites as defined by ENCODE²² for cells potentially relevant to lung function. We also carried out conditional analyses, using GCTA⁶⁵, of sentinel variants conditioning on functional variants within the loci to assess whether the association signals were explained by functional variants (P value of the sentinel variant conditioned on the functional variant, conditional P , > 0.01). Functional variants were defined using SIFT⁶⁶, PolyPhen-2 (ref. 67), CADD⁶⁸ and GWAVA⁶⁹ databases. Additional analyses were undertaken for a subset of priority genes within the 16 loci (description of the selection is given in the Supplementary Methods). These included RNA-seq analyses to confirm messenger RNA expression in a lung-relevant cell (bronchial epithelium) and detect novel splice isoforms; assessment of differential expression across pseudoglandular and canalicular stages of human fetal lung development using gestational age as a continuous variable in linear regression²⁵, and assessment of differences in expression levels in bronchial brushings between COPD cases and smoking controls⁷⁰. Details for all these analyses are provided in the Supplementary Methods.

Associations with other traits. The association of the 16 sentinel variants with the following traits was assessed: lung function in children undertaking the same analysis as for adults in the ALSPAC data set⁴²; gene by smoking interaction by undertaking a Z-test comparing the effect of a given variant in ever smokers and in never smokers using stage-1 results; smoking behaviour by undertaking a logistic regression analysis with heavy- versus never-smoking status as an outcome in the UK BiLEVE data set. In addition, the GWAS catalog⁴³ was queried for variants in 2-Mb regions centred on the sentinel variant for the 16 loci. Variants that were genome-wide significant ($P < 5 \times 10^{-8}$) in the GWAS catalog⁴³ and were in LD ($r^2 > 0.3$) with the sentinel variants, or were in genes that contained at least one variant in LD ($r^2 > 0.3$) with the sentinel variants were selected.

Pathway analyses. Stage-1 GWAS results were tested for enrichment of known biological pathways using MAGENTA v2 (ref. 52). Six databases of biological pathways, including Ingenuity Pathway (June 2008, number of pathways $n = 81$), KEGG (2010, $n = 186$), PANTHER Molecular Function (January 2010, $n = 216$), PANTHER Biological Processes (January 2010, $n = 217$), PANTHER Pathways (January 2010, $n = 94$) and Gene Ontology (April 2010, $n = 1778$), were tested. An FDR threshold of 5% was used and significance thresholds were Bonferroni corrected for each database. Genes within 500 kb either side from the sentinel variants were flagged in the analysis. Sensitivity analyses were run after removing genes in the human leukocyte antigen region on chromosome 6. More details on the method are provided in the Supplementary Methods.

Additional analyses. Heterogeneity tests were undertaken for the 16 sentinel variants in stage 1. We undertook stepwise conditional analyses as performed by GCTA⁶⁵ in each locus to identify additional signals. Full methods and results are described in the Supplementary Notes.

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