Applying whole-genome sequencing in relation to phenotype and outcomes in siblings with cystic fibrosis

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Abstract Variations in disease onset and/or severity have often been observed in siblings with cystic fibrosis (CF), despite the same CFTR genotype and environment. We postulated that genomic variation (modifier and/or pharmacogenomic variants) might explain these clinical discordances. From a cohort of patients included in the Wisconsin randomized clinical trial (RCT) of newborn screening (NBS) for CF, we identified two brothers who showed discordant lung disease courses as children, with one milder and the other more severe than average, and a third, eldest brother, who also has severe lung disease. Leukocytes were harvested as the source of DNA, and whole-genome sequencing (WGS) was performed. Variants were identified and analyzed using in-house-developed informatics tools. Lung disease onset and severity were quantitatively different between brothers during childhood. The youngest, less severely affected brother is homozygous for HFE p.H63D. He also has a very rare PLG p.D238N variant that may influence host–pathogen interaction during chronic lung infection. Other variants of interest were found differentially between the siblings. Pharmacogenomics findings were consistent with the middle, most severely affected brother having poor outcomes to common CF treatments. We conclude that genomic variation between siblings with CF is expected. Variable lung disease severity may be associated with differences acting as genetic modifiers and/or pharmacogenomic factors, but large cohort studies are needed to assess this hypothesis.

[Supplemental material is available for this article.]

INTRODUCTION

Variations in disease onset and/or severity have been observed in siblings with cystic fibrosis (CF) for decades (Macgregor and Rhaney 1948). The most notable differences relate to CF lung disease. Many factors have been suggested to explain discordances such as gender, meconium ileus, age of diagnosis, cystic fibrosis transmembrane conductance regulator (CFTR) genotype, acquisition of respiratory pathogens, treatment regimens, and, most recently, genetic modifiers (Orenstein et al. 1977; Kerem and Kerem 1996; Li et al. 2004; Cutting 2010; Sanders et al. 2014; O’Neal and Knowles 2018).

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Ontology terms: bronchiectasis; chronic lung disease; diabetes mellitus; elevated sweat chloride; exocrine pancreatic insufficiency; meconium ileus; multiple bilateral pneumothoraces; recurrent Aspergillus infections; recurrent respiratory infections

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Prior to the onset of routine early diagnosis through newborn screening (NBS), it was proposed from an influential sibling pair study that diagnosis at a younger age leading to early treatment initiation might account for lessened severity in the younger sibling (Orenstein et al. 1977). Although NBS does lead to a reduction in the severity of outcomes for the majority of children (Farrell et al. 2001), there is a subgroup known as “nonresponders,” approximately one-third of patients, who do not show the same degree of benefit from early diagnosis (Sanders et al. 2018). After discovery of the CFTR gene (Kerem et al. 1989), studies on genotype–phenotype relationships, although illuminating with regard to pancreatic functional variations (Kerem and Kerem 1996; Durno et al. 2002), did not succeed in explaining differences in the onset and severity of lung disease (Kerem et al. 1990; Kerem and Kerem 1996). In fact, observations on homozygous F508del patients revealed such great variability in respiratory disease outcomes that attention turned to the identification of genetic modifiers (Cutting 2010; Corvol et al. 2015; O’Neal and Knowles 2018). Application of genome-wide association studies (GWASs) has more recently facilitated identification of certain classes of modifiers resulting in phenotypic variation (Corvol et al. 2015).

Individual specific variable drug responses can present significant challenges for the effective and safe treatment of many human diseases. Pharmacogenomics (PGx) attempts to relate drug response to patient-specific genome in order to better customize therapeutic selections. We posited that the whole-genome sequencing (WGS) data generated for phenotypic modifier studies might simultaneously be useful for the extraction of PGx findings of relevance for patients with CF.

Here we report an investigation of non-CFTR genetic factors in three brothers with CF who showed considerable clinical variations in the course of disease manifestations, particularly during childhood. Their discordances led us to search for genetic modifiers and pharmacogenomics differences using a WGS-based approach. Our hypothesis is that genomic variation will help explain clinical differences among siblings.

**RESULTS**

**Patient Phenotypes**

All three brothers were confirmed by WGS to have a CFTR genotype of F508del/CFTRdele2,3 and no other CFTR variants. None had meconium ileus, each was diagnosed with a sweat chloride level of >100 mmol/L, and all had pancreatic insufficiency. Their clinical histories are summarized in Table 1. Quantitative longitudinal evaluations of lung disease were performed and data are shown in Figures 1 and 2.

The eldest brother was diagnosed with CF at 3 mo of age because of failure to thrive. Despite initiation of pancreatic enzyme replacement therapy (PERT), he tracked under the fifth percentile for weight throughout childhood. After 12 yr of *Pseudomonas aeruginosa* (PA)-positive cultures and progressive lung disease, his sputum culture revealed *Aspergillus fumigatus* (AF) at 18 yr. He received a bilateral lung transplantation at 29 yr of age after his percent predicted forced expiratory volume in 1 sec (ppFEV1) had decreased to 20% and his Wisconsin chest X-ray (WCXR) score was rising dramatically (Fig. 2).

The middle brother was diagnosed with CF through NBS and began PERT at 36 wk of age when pancreatic insufficiency was demonstrated through protocol-guided studies (Farrell 2000). Although PA was not cultured until 16 yr of age, he was noted by WCXR scores to have developed irreversible lung disease (i.e., WCXR > 5) (Farrell et al. 2003) by 2 yr of age and had severely progressive lung disease between 5 and 14 yr (Fig. 2). His sputum culture first grew AF at age 11 yr. He was treated for AF with itraconazole, but eradication was unsuccessful. He underwent bilateral lung transplantation age 20 yr, and died at 24 yr because of multisystem organ failure.
Figure 1. Pulmonary function test results in the three brothers (i.e., percent predicted FEV1 [ppFEV1] values as related to age). The data obtained on three brothers are shown as A (eldest), B (middle), and C (youngest) and demonstrated that the youngest had better pulmonary function than his brothers. After 20 yr, the eldest brother deteriorated and reached 20% at 28 yr of age.
The youngest brother was also diagnosed with CF through NBS. He had a relatively mild course of lung disease throughout most of childhood; this was clearly less severe than the cohort of other screened patients in the RCT (Fig. 2). After age 12 yr, he began to deteriorate in association with culturing AF. He has yet to acquire PA through his current age of 26 yr.

Modifiers of Cystic Fibrosis

Our WGS-based analysis identified five variants that might affect the brothers’ phenotypes (Table 2). All three brothers were found to have a common heterozygous nonsynonymous variant in \( EPHX1 \) previously shown to reduce enzyme activity by at least 20% (Hassett et al. 1994). A low activity of \( EPHX1 \) has been identified as a risk factor for chronic obstructive pulmonary disease (COPD), and this specific variant has been associated with altered risk of pulmonary disease (Smith and Harrison 1997; Vibhuti et al. 2007; Castaldi et al. 2009; Carlton et al. 2003). Although association between variant homozygosity and increased risk of COPD was recently confirmed via comprehensive meta-analysis, the findings were not supported for variant heterozygosity (Li et al. 2013).

A known deleterious missense variant in \( HFE \) (p.H63D) was found to be heterozygous in the eldest and middle brothers and homozygous in the youngest. This variant is pathogenic for a mild form of hereditary hemochromatosis (Feder et al. 1996). Although clinical hemochromatosis is mostly associated with p.C282Y homozygosity, there is evidence that p.H63D can act as a modifier for α1-antitrypsin deficiency liver and lung disease (Joly et al. 2017). Functional variation in \( HFE \) has been associated with risk of infection in general (Khan et al. 2007; Nairz et al. 2017; Gill et al. 2019) and deteriorating lung function specifically in CF (Smith et al. 2019).
A third variant of interest is the known deleterious variant in \textit{MBL2} (D variant; p.R52C), which was heterozygous in the eldest and youngest brothers. \textit{MBL2} plays a role in immune function, activating the lectin complement pathway and leading to clearance by phagocytes. This variant is pathogenic for mannose-binding protein deficiency (Heitzeneder et al. 2012). Variants in this gene have been associated with recurrent respiratory tract infections, bronchiectasis severity, and age-related decline in lung function (Horiuchi et al. 2005; Haerynck et al. 2012; Liu et al. 2012; Pifferi et al. 2015). The presence of functional variants in \textit{MBL2} has shown association with different aspects (lung function, infection, and survival) of CF in some studies (Dorfman et al. 2008; McDougal et al. 2010).

The youngest brother was heterozygous for an ultra-rare, conserved, nonsynonymous predicted to be damaging variant (p.D238N) in a known kringle-binding domain in the plasminogen protein (\textit{PLG}) (Arai et al. 1998). Although not previously reported, this variant would be categorized as likely pathogenic based on American College of Medical Genetics and Genomics (ACMG) guidelines (Richards et al. 2015). Functional variation in \textit{PLG} is associated with plasminogen deficiency, which has been causally linked to a rare chronic inflammatory disease of the mucous membranes. PLG deficiency also predisposes patients to recurrent sinobronchial AF infections in immunocompromised individuals (Zaas et al. 2008).

Finally, a known deleterious variant in \textit{SERPINA1} (PiS; p.E288V) was identified in the middle and youngest brothers. This variant is a known risk factor for COPD and a known pathogenic variant for \textit{α1}-antitrypsin deficiency, which increases the risk for pulmonary and liver disease in both heterozygous and homozygous states (Pérez-Rubio et al. 2015; Greene et al. 2016). Other variants in this gene have been shown to modify CF-related disease (Debray et al. 2019).
Pharmacogenomic Findings

Pharmacogenomic variants categorized as level 1A, 1B, 2A, and 2B (Whirl-Carrillo et al. 2012) were identified for a number of drugs commonly used in the treatment of CF (Table 3). Whereas all three brothers are similar for a number of loci, there remain some notable differences. For example, several loci assessed (CRHR2, COL22A1, and ADRB2) have been

| Drugs       | Eldest brother | Middle brother | Youngest brother |
|-------------|----------------|----------------|------------------|
| Tobramycin  | MT-RNR1—patient is homoplasmic (A) for this rs267606617 allele. | MT-RNR1—patient is homoplasmic (A) for this rs267606617 allele. | MT-RNR1—patient is homoplasmic (A) for this rs267606617 allele. |
|            | Associated with decreased risk of aminoglycoside-induced hearing loss. | Associated with decreased risk of aminoglycoside-induced hearing loss. | Associated with decreased risk of aminoglycoside-induced hearing loss. |
| Salbutamol  | CRHR2—patient is homoyzous (A/A) for the rs7793837 allele; associated with a better response (increased acute bronchodilation). | CRHR2—patient is heterozygous (A/T) for the rs7793837 allele; associated with intermediate response. | CRHR2—patient is homozygous (A/A) for the rs7793837 allele; associated with a better response (increased acute bronchodilation). |
|            | COL22A1—patient is homozygous (C/C) for the rs6988229 allele; associated with a decreased bronchodilator response (FEV1). | COL22A1—patient is homozygous (C/C) for the rs6988229 allele; associated with a decreased bronchodilator response (FEV1). | COL22A1—patient is homozygous (C/C) for the rs6988229 allele; associated with a decreased bronchodilator response (FEV1). |
|            | ADRB2—patient is heterozygous (A/G) for the rs1042713 allele. Children with the AA genotype with asthma who are treated with salmeterol or salbutamol may have a decreased response to treatment than children with G/G genotypes, and better response than children with A/A genotype. | ADRB2—patient is homozygous (A/A) for the rs1042713 allele. Children with the AA genotype with asthma who are treated with salmeterol or salbutamol may have a decreased response to treatment than children with G/G genotypes, and better response than children with A/A genotype. | ADRB2—patient is homozygous (A/A) for the rs1042713 allele. Children with the AA genotype with asthma who are treated with salmeterol or salbutamol may have a decreased response to treatment than children with G/G genotypes, and better response than children with A/A genotype. |
| Omeprazole  | CYP2C19 1B/*2B—intermediate-to-poor metabolizer. Patient may require lower doses of drugs metabolized by CYP2C19 enzyme compared to normal/rapid metabolizers to avoid toxicity or switching to a drug independent of CYP2C19 metabolism. | CYP2C19 1B/*2B—intermediate-to-poor metabolizer. Patient may require lower doses of drugs metabolized by CYP2C19 enzyme compared to normal/rapid metabolizers to avoid toxicity or switching to a drug independent of CYP2C19 metabolism. | CYP2C19 1B/*2B—intermediate-to-poor metabolizer. Patient may require lower doses of drugs metabolized by CYP2C19 enzyme compared to normal/rapid metabolizers to avoid toxicity or switching to a drug independent of CYP2C19 metabolism. |
|            | CYP3A5 (*3/*3)—patient is homozygous for the two alleles (rs776746 and rs28383479) that make up the CYP3A5*3 haplotype. Decreased metabolizer; may require a lower dose. | CYP3A5 (*3/*3)—patient is homozygous for the two alleles (rs776746 and rs28383479) that make up the CYP3A5*3 haplotype. Decreased metabolizer; may require a lower dose. | CYP3A5 (*3/*3)—patient is homozygous for the two alleles (rs776746 and rs28383479) that make up the CYP3A5*3 haplotype. Decreased metabolizer; may require a lower dose. |
| Tacrolimus  | TPMT (*1A/*1A)—associated with (1) increased inactivation of thiopurines due to normal TPMT activity; (2) an increased risk for toxicity when receiving thiopurine drugs and purine analogs. Patients may still be at risk for toxicity when taking thiopurine drugs and purine analogs. | TPMT (*1/*3A)—associated with (1) decreased inactivation of thiopurines due to deficient TPMT activity; (2) an increased risk for toxicity when receiving thiopurine drugs and purine analogs. | TPMT (*1A/*1A)—associated with (1) increased inactivation of thiopurines due to normal TPMT activity; (2) a decreased risk for toxicity when receiving thiopurine drugs and purine analogs. Patients may still be at risk for toxicity when taking thiopurine drugs and purine analogs. |

*Prescribing information based on CYP2C19 haplotypes is available for a number of other drugs, including the following: amitriptyline, citalopram, escitalopram, esomeprazole, clomipramine, clopidogrel, doxepin, imipramine, lansoprazole, moclobemide, rabeprazole, sertraline, and trimipramine.
associated with efficacy of the β2-adrenergic salbutamol (Bleecker et al. 2007; Poon et al. 2008; Duan et al. 2014), with each brother having unique combinations. Of these, the middle brother might have the least benefit from albuterol. Similarly, the middle brother was unique in his TMPT haplotype associated with thiopurine (e.g., azathioprine) inactivation that would put him at higher risk of toxicity when treated with thiopurine drugs.

CYP2C19 is a highly polymorphic enzyme that contributes to the metabolism of many clinically relevant drugs (Wrighton et al. 1996; Gaedigk et al. 2018). Individuals can be categorized as ultra-rapid, rapid, intermediate, or poor metabolizers, depending on absent, decreased, normal, or increased function (Scott et al. 2013). As shown in Table 3, we were able to determine that all three brothers had a no function copy of this gene (*2B) and that both the eldest and youngest brothers have a *1B second copy, whereas the middle brother’s second allele is *17. Thus, at best, the middle brother is an intermediate metabolizer, because the *17 gain-of-function allele does not completely compensate for the no function allele.

Multiple genes encode CYP3A enzymes, with CYP3A4 and CYP3A5 regarded as the most important (Williams et al. 2002). Expression is highly polymorphic with 52 CYP3A4 and 26 CYP3A5 star alleles cataloged (Gaedigk et al. 2018). All of the brothers were found to be homozygous for the two alleles that make up the CYP3A5*3 haplotype (CYP3A5*3/3) and are thus considered to be CYP3A5 non-expressors who may have decreased metabolism of a large percentage of oxidatively metabolized drugs (Williams et al. 2002). This variation in expression is particularly important for drugs and substrates with narrow therapeutic indices, including the immunosuppressants cyclosporine and tacrolimus, and can result in clinically significant differences in toxicity and response (Ericson et al. 2017). Additional PGx related findings can be found in Supplemental Table 1.

DISCUSSION

O’Neal and Knowles (2018) stated that “[t]he not-too-distant future holds many more opportunities for discovery of CF gene modifiers … whole-genome sequencing will expand GWAS-type studies to rare variant analyses.” In support of their prediction, we suggest that our WGS assessment of the three brothers identified several variants that could predispose a patient to better or worse outcomes and that the combination of exploring for genetic modifiers plus pharmacogenomics variants could be valuable. It was particularly interesting to compare the mildly affected youngest brother with his siblings and observe notable differences in both potential genetic modifiers and pharmacogenomics factors.

We suggest that the variants identified in these brothers may contribute to their varying courses of CF lung disease, including a known disease-associated variant in SERPINA 1 (middle and youngest), a known CF modifier (Bartlett et al. 2009), and the HFE p.H63D variant found in a heterozygous state in the middle and eldest brother and in a homozygous state in the youngest brother. The former is known to increase risk of liver disease in CF (Bartlett et al. 2009) and COPD and liver disease in otherwise healthy individuals (Pérez-Rubio et al. 2015; Greene et al. 2016). HFE variants have been reported to be present at a surprisingly high frequency among CF patients (Devaney et al. 2003; Smith et al. 2019) and appear to be modifiers of lung disease severity based on accelerating pulmonary dysfunction (Smith et al. 2019). This effect might be attributable to altering susceptibility to infections with many bacteria, including PA (Khan et al. 2007).

The known deleterious MBL2 p.R52C variant, which was heterozygous in the eldest and youngest brothers, has been associated with recurrent respiratory tract infections and greater bronchiectasis severity in patients with ciliary dyskinesia (Pifferi et al. 2015). It has also been reported that mannose-binding lectin (MBL)-deficient CF patients are colonized with PA earlier compared to MBL-sufficient patients (Haerynck et al. 2012; Gravina et al. 2015),
and MBL has been linked to the age-related decline in lung function in CF patients (Garred et al. 1999; Pifferi et al. 2015). Consistent with these data, the eldest brother colonized PA at the youngest age.

Only the youngest brother was found to be heterozygous for a very rare nonsynonymous variant in the plasminogen gene (PLG; p.D238N). Plasminogen has been shown to directly bind to AF and individuals with PLG deficiency can have tracheobronchial fibrin deposits, which have been shown to impair the ciliary system, thereby predisposing to recurrent sino-bronchial infections (Schuster et al. 2007). In a patient with CF in which mucociliary clearance is already compromised (Chatelin et al. 2017), PLG deficiency may increase the risk of *Aspergillus* bronchitis or related symptoms (Zaas et al. 2008). The presence of the variant in the youngest brother is consistent with his deterioration after 12 yr of age, when he showed recurrent AF-positive cultures.

*CYP2C19* status explains ∼40% of the variability in response to azole antifungal agents for systemic use (Berge et al. 2011). Given the potential severity of illness in CF patients with invasive fungal infections, it is critical that therapeutic antifungal concentrations be attained rapidly (Berge et al. 2011; Hamadeh et al. 2017). Variability in response complicates optimal dosing and can increase (dependent on genotype) either risk of toxicity or risk of undertreatment (Berge et al. 2011; Hamadeh et al. 2017). Of note, in this case, the youngest brother, who may be at increased risk for AF infection owing to modifiers, may also be at increased risk of toxicity at standard therapeutic antifungal dosage because of *CYP2C19* genotype. Although a lower dose may be necessary to avoid toxicity in such individuals, it is important to note that the severity of fungal infection–related disease combined with high variability in antifungal response already necessitates a therapeutic drug monitoring (TDM) approach that would be expected to dramatically reduce toxicity associated risk in this individual (Billaud et al. 2010).

**Conclusions**

Our data show that there is notable genomic variation in CF related loci across these three siblings with CF. A complex interplay between the risk to disease or infection conferred by genetic modifiers and the response to drugs used to treat the disease and infection as the result of pharmacogenomics were likely contributory to the clinical outcomes in these patients.

To our knowledge, this is the first application of WGS to individual patients with CF applied in an effort to clarify (and eventually predict) the course of the disease in a strategy of personalized/precision medicine. However, additional cohort studies are needed to further assess this hypothesis. We also suggest that analysis of WGS data initially generated for identification of causal or modifier variation might also provide useful for the generation of pharmacogenomics data. Clearly, further evaluation with regards to specificity, sensitivity, and reproducibility would be required.

**METHODS**

**Summary of RCT Protocol**

The Wisconsin RCT of NBS was initiated in April 1985, and patients were eligible for enrollment until July 1994. The protocol for this unique project was conducted before mandatory registration and has been described in detail elsewhere (Farrell et al. 2001). Relevant to this study, each patient was managed with an Evaluation and Treatment Protocol (Farrell 2000) and had clinical outcome variables assessed systematically every 3 mo, along with regular nutritional status evaluations using anthropometry and biochemical markers. To evaluate lung disease, we used (Farrell et al. 2003) a combination of spirometry and longitudinal
quantitative radiology annually and biannually, respectively, scoring the chest radiographs with the sensitive WCXR scoring method—a rating system that scores the severity of air trapping, peribronchial thickening, bronchiectasis, other indicators of infection, and atelectasis (Koscik et al. 2000). From this cohort of patients, we identified for WGS two brothers who as children showed lung disease course discordances—one, the youngest, was milder, whereas the middle brother appeared more severe than average—and a third, eldest brother (born before the RCT) also had severe lung disease.

Sample Handling and Whole-Genome Sequencing
After we obtained IRB approvals from the University of Wisconsin Human Subjects Committee and written informed consent obtained directly for research and publication, we collected whole blood samples from each brother in 4 mL volumes, anticoagulated specimens with EDTA, and then harvested leukocytes from the postcentrifugation buffy coats. The middle brother’s DNA was isolated from the blood sample at the University of Wisconsin, where the sample underwent WGS using an Illumina HiSeq 4000 instrument. The harvested leukocyte samples from the eldest and youngest brothers were sent frozen to HudsonAlpha Institute for Biotechnology for on-site DNA isolation using the QIAasympohony DSP DNA Mini Kit. The extracted DNA was then quantified using a Qubit system as well as using agarose gel electrophoresis to ensure sufficient yield and quality for sequencing. The eldest and youngest brothers’ DNA samples were sequenced at HudsonAlpha Institute for Biotechnology’s Genomic Services Laboratory using an Illumina HiSeqX paired-end 150-bp read methodology. Samples were sequenced to a mean depth of 35× coverage with >80% of bases covered at 20×. Specific coverage information for variants reported is provided in Table 2.

Secondary and Tertiary Analyses
Secondary and tertiary analyses followed standard methods (Worthey 2017). In brief, sequences for each of the brothers were aligned to reference genome hg19 using BWA-mem according to standard protocols (Li and Durbin 2009). The aligned sequences were processed using the GATK package for base quality score recalibration, insertion/deletion realignments, duplicates removal, and discovery of single-nucleotide substitutions and small insertions and deletions, as well as genotyping (DePristo et al. 2011). Quality control metrics and sample swap tests were applied, including blinded CFTR genotype determinations and sample relationship metrics. The resultant variant files were loaded into the tertiary analysis software, Codicem (http://envisiongenomics.com/codicem-analysis-platform/), developed by the Worthey Laboratory at the HudsonAlpha Institute for Biotechnology. Coverage information for variants reported is provided in Table 2.

Codicem was used to analyze and support interpretation of the variant data (described elsewhere (Holt et al. 2019). Preset filter steps were used to remove variants of poor quality as well as variants in genes on the American College of Medical Genetics list of 59 genetic conditions with no phenotype connection to CF (Kalia et al. 2017). This was used to reduce, if not eliminate, the potential for identifying incidental findings that should be returned, per the IRB-approved protocol. The filtering strategy used in Codicem also prioritized and excluded variants based on various in silico damage predictions, existing disease associations including CF modifiers (e.g., Corvol et al. 2015), phenotype terms, and population allele frequencies. This reduced the number of variants to be reviewed to those most likely to be significant. Variants passing these filters were reviewed by two variant analysts who further assessed variants based on their potential effect on gene function and CF pathogenesis/progression. Variants of interest were selected and passed to team review. Variants subsequently deemed likely contributory underwent Sanger confirmation. These analyses allowed us to create the list of potential genetic modifiers provided in Supplemental Table 2.
Pharmacogenomics Analysis
Pharmacogenomic definitions were downloaded from PharmGKB (Barbarino et al. 2018), an expert-curated database, in order to build a pharmacogenomics database containing simple (i.e., single-variant) to complex (i.e., many-variant) haplotype/diplotype associations along with drug metabolism, response, and adverse reaction data. The association of pharmacogenomic findings for a given patient followed a three-step process. First, the pharmacogenomics annotation pipeline, an in-house application written in Java, gathered all relevant variant information for a patient from a variant call format (VCF) file. Next, diplotypes and genotypes were inferred using this data. Finally, pharmacologically relevant curations for each patient were extracted by matching haplotype to association data. All curations selected to be of interest were manually rereviewed by two analysts. Passing curations were categorized by drug class, as defined by Drug Bank (Wishart et al. 2018). Variants for drug classes of interest (e.g., antibiotics and immunosuppressants) were manually reviewed to identify potentially significant variant-drug combinations. Pharmacogenomic variants present in each of the brothers were automatically annotated with their respective effects matched to this database and subject to additional manual curation.

ADDITIONAL INFORMATION

Data Deposition and Access
Deidentified variant data has been submitted to ClinVar (https://www.ncbi.nlm.nih.gov/clinvar/) under accession numbers SCV001142519.1, SCV001142520.1, SCV001142521.1, SCV001142522.1, and SCV001142523.1. Raw sequencing data could not be deposited because of lack of consent.

Ethics Statement
This study was approved by the IRB at the University of Wisconsin. Appropriate written consent was obtained prior to participation in this study.

Author Contributions
E.A.W. and P.M.F. were responsible for the conception and design of this study and supervised the work. A.T.B., P.M.F., and A.L. were involved in gathering of clinical data and patient management. M.A.W., C.B., D.B., J.M.H., N.S., B.W., and E.A.W. contributed to genomic analysis and interpretation. M.A.W., C.B., D.B., A.T.B., N.S., B.W., P.M.F., and E.A.W. contributed to writing this work, and all authors participated in reviewing the manuscript.

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