When splicing is not all or none: Implications for variant classification

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Abstract

Variants with intermediate functional effects—those that neither abolish nor fully preserve gene

function—constitute an underappreciated source of genetic complexity and a grey zone for

classification within the current ACMG/AMP binary pathogenic-or-benign framework. Here, we

address this issue using GT>GC (+2T>C) 5′ splice site variants, approximately 15% of which retain

variable amounts of wild-type transcript, as a tractable model. Drawing on disease-associated GT>GC

variants and those analyzed by full-length gene splicing assays, we show that residual wild-type

transcript levels form a continuum from barely detectable to near normal. Notably, GT>GC variants

retaining substantial wild-type transcript (e.g., *SPINK1* c.194+2T>C, *HBB* c.315+2T>C, and *BRCA2*

c.8331+2T>C) frequently present classification challenges. Our findings also indicate that in classical

Mendelian disorders caused by monoallelic or biallelic complete loss-of-function alleles, GT>GC

variants are not invariably pathogenic (e.g., *DOCK8* c.3234+2T>C). Minigene analyses of selected

CFTR and *BAP1* variants further illustrate the inherent difficulty of predicting and experimentally

quantifying residual wild-type transcript levels (e.g., *CFTR* c.2490+2T>C and *CFTR* c.4242+2T>C).

Because GT>GC variants that generate wild-type transcript exemplify the degeneracy and context

dependence of splice-site recognition, insights from their study extend to other splice-altering variants

outside the invariant +1, −1, and −2 positions. Overall, our analysis highlights that grey-zone variants

with intermediate functional effects represent a biological reality that complicates the genotype–

phenotype relationship and underscores the need for classification frameworks that accommodate the

full spectrum of variant effects and the true complexity of human genetic disease.

Keywords

39 ACMG/AMP variant classification guidelines; ClinVar; Incomplete penetrance; Intermediate

40 functional effects; Loss-of-function (LoF) variants; Missing heritability; Residual wild-type transcript;

41 SpliceAI; Variant effect spectrum; Variants of uncertain significance (VUS)

Introduction

Accurate classification of genetic variants remains a central challenge in human genetics.[1,2] The

increasing use of a genotype-first or genomic-first approach,[3,4] enabled by exome and genome

sequencing, has greatly magnified this challenge. In many cases, the mode of inheritance becomes

apparent only after genetic analysis, and disorders presumed to follow a simple Mendelian pattern may

instead have a complex, multifactorial, or oligogenic basis. This complexity underscores the

limitations of the widely used American College of Medical Genetics and Genomics and the

Association for Molecular Pathology (ACMG/AMP) framework.[5] Its dichotomization of variants as

either "pathogenic" or "benign," although useful for clinical consistency, cannot account for the full

spectrum of variant effects that contribute to human genetic disease.[3,6-9] It should be noted that the

current ACMG/AMP category "variant of uncertain significance (VUS)"[5] is not a biological entity but

a temporary designation expected to resolve as either pathogenic or benign once sufficient evidence

becomes available.

For clarity, we focus here on variants that act through a loss-of-function (LoF) mechanism. In

contrast to variants at the two extremes—those causing complete or near-complete loss of function of

the affected allele, typically responsible for classic Mendelian disease phenotypes, and those retaining

complete or near-complete normal function—variants with intermediate effects are challenging to

interpret because they lie outside the scope of binary pathogenicity models. This challenge is

illustrated by the multiple partial LoF variants identified in *SPINK1* (MIM: 167790).[8] Despite their

clear biological reality and clinical relevance, such variants remain underexplored and are not

adequately addressed by any widely established classification frameworks, which offer limited

guidance for interpreting quantitative or intermediate functional effects and thereby risk systematic

over- or under-classification.

In this study, we explore GT>GC (+2T>C) 5′ splice site variants as a tractable model to investigate

how intermediate functional effects challenge classification. Although substitutions at canonical GT

donor or AG acceptor sites typically abolish normal splicing,[10,11] a subset of GT>GC substitutions

defy this expectation. Specifically, a meta-analysis of 45 disease-associated variants, together with

full-length gene splicing assay (FLGSA) data from 103 substitutions, showed that 15–18% retain

residual wild-type (WT) transcript, in some cases at levels approaching 80% of normal.[12] This

variability can be clinically significant. For instance, in cystic fibrosis, as little as 5% normal cystic

fibrosis transmembrane conductance regulator (CFTR) activity may be sufficient to prevent

pulmonary disease,[13,14] whereas in the hemophilias, factor VIII or IX levels above 5% may lead to

only mild bleeding diatheses.[15,16] Conversely, in disorders typically caused by complete LoF alleles,

GT>GC variants may not be invariably pathogenic if substantial WT transcript is preserved, as we

previously proposed.[12]

Using residual (leaky) WT transcript levels as a quantitative measure, this study aims to identify

the conceptual and methodological limitations in classifying GT>GC variants that generate substantial

yet incomplete amounts of WT transcript. Our analysis highlights the classification challenges posed

by intermediate functional effects and reinforces the need for expanded frameworks capable of

capturing the full spectrum of variant effects.[6,8]

Methods

Identification and selection of GT>GC variants

Our 2019 study[12] served as the starting point. From the FLGSA-analyzed GT>GC variants reported

therein, we included all variants that both retained WT transcript and were registered in ClinVar

(https://www.ncbi.nlm.nih.gov/clinvar/; accessed September 5, 2025) for analysis. To identify

additional disease-associated GT>GC variants with evidence of residual WT transcript generation

published since 2019, we conducted a Google-based literature search using the keyword "+2T>C,"

complemented by cascade reference tracking and cross-referencing within the peer-reviewed literature

(up to September 5, 2025). As in our previous work,[12] we focused on studies in which RNA analysis

was performed on material derived from affected individuals, and simple heterozygotes were excluded

unless residual WT transcript generation from the variant allele was explicitly demonstrated by allele-

specific reverse transcription-PCR (RT-PCR) in the original study. At the single-gene level, all 26

possible *CFTR* (MIM: 602421) GT>GC variants were subjected to initial evaluation. Finally, a *BAP1*

(MIM: 603089) variant, c.783+2T>C (GenBank: NM_004656.4, NG_031859.1), identified during the

literature search as having conflicting classifications,[17,18] was included for additional minigene splicing

analysis.

In silico splicing predictions

102 *In silico* predictions were obtained using SpliceAI[19] via the SpliceAI Lookup tool

103 (https://spliceailookup.broadinstitute.org/; accessed September 5, 2025) with the following settings:

104 genome version, hg38; GENCODE, basic; maximum distance, 10,000 bp; and REF & ALT scores

105 displayed. The resulting delta donor loss (ΔDL) scores, derived from comparisons between WT and

106 variant sequences, were used for analysis.

Variant classification or description in public databases

Selected variants were queried in ClinVar (https://www.ncbi.nlm.nih.gov/clinvar/)[20] to determine their

current classification status. The Clinical and Functional TRanslation of CFTR database (CFTR2;

http://www.cftr2.org/)[21] and the CFTR-France database (https://cftr.chu-montpellier.fr/)[22] were

consulted to evaluate *CFTR* variant classifications. Variant information from the Cystic Fibrosis

Mutation Database (CFTR1; http://www.genet.sickkids.on.ca/) was used to assist in the classification

of two *CFTR* variants that were not registered in either CFTR2 or CFTR-France. Our evaluation of the

variant classification or description data in these databases was last updated on October 28, 2025.

Minigene splicing analysis of selected *CFTR* GT>GC variants

Variant selection

Of the 26 possible *CFTR* GT>GC variants (Table S1), we selected all three with low SpliceAI DL

scores (≤0.56)—c.2490+2T>C, c.3717+2T>C and c.4242+2T>C—together with one variant

(c.3873+2T>C) that had a high SpliceAI score (0.98) but was previously reported to generate WT

transcript,[23] for minigene splicing analysis.

NM_000492.4 and NG_016465.4 were used as the mRNA and genomic reference sequences of

CFTR, respectively.

Construction of pET01 and pSPL3 minigene expression vectors

For each selected +2T>C variant, the corresponding WT genomic sequences cloned into the pET01

and pSPL3 exon trapping vectors[24] were always identical. Each *CFTR* WT insert comprised 213–217

bp from the 3′ end of intron N−1, the entire exon N, and 213–216 bp from the 5′ end of intron N (N =

variant-affected intron). Cloning of WT constructs and introduction of the selected +2T>C variants,

followed by Sanger confirmation, were performed by GENEWIZ Biotech Co. (Suzhou, China).

Cell culture, transfection, RT-PCR, and sequencing

Experiments were carried out essentially as previously described.[24] Human embryonic kidney 293T

(HEK293T) cells were maintained in DMEM (Gibco) supplemented with 10% fetal calf serum

135 (Procell). Cells (3.5×10^5/well) were seeded in 6-well plates 24 h before transfection. Per well, 2.5 μg

136 WT or variant plasmid were transfected using HieffTrans Universal Transfection Reagent (Yeasen).

137 Forty-eight hours later, total RNA was extracted with the FastPure Cell/Tissue Total RNA Isolation

138 Kit V2 (+gDNA wiper) (Vazyme). RT was carried out using the HiScript III 1st Strand cDNA

139 Synthesis Kit (Vazyme), incorporating 2 μL 5×gDNA wiper Mix, 2 μL 10×RT Mix, 2 μL HiScript III

140 Enzyme Mix, 1 μL Oligo (dT)20VN, and 1 μg total RNA. RT-PCR was performed in a 25-μL reaction

141 mixture containing 12.5 μL 2×Taq Master Mix (Vazyme), 1 μL cDNA, and 0.4 μM each primer. The

142 primer pairs were as follows: pET01 vector—forward 5′-GAGGGATCCGCTTCCTGGCCC-3′,

143 reverse 5′-CTCCCGGGCCACCTCCAGTGCC-3′; pSPL3 vector—forward 5′-

144 TCTGAGTCACCTGGACAACC-3′, reverse 5′-ATCTCAGTGGTATTTGTGAGC-3′. The PCR

145 program comprised an initial denaturation step at 94°C for 5 min, followed by 35 cycles of

146 denaturation at 94°C for 30 s, annealing at 55°C for 30 s, and extension at 72°C for 1 min, and a final

147 extension step at 72°C for 7 min. RT-PCR products were separated on agarose gels, excised, and

148 purified (Omega Bio-Tek Gel Extraction Kit). Sequencing primers were the same as those used for

149 RT-PCR, and reactions were performed with the BigDye Terminator v3.1 Cycle Sequencing Kit

150 (Applied Biosystems).

152 **Full-length and minigene splicing analysis of *BAP1* c.783+2T>C**

153 The full-length genomic *BAP1* WT insert, spanning from the translation initiation codon to the

154 termination codon, was amplified from genomic DNA of a normal control using long-range PCR. The

155 resulting fragment was cloned into the pcDNA3.1/V5-His-TOPO vector. These procedures, as well as

156 subsequent transfection and RT–PCR, were performed essentially as previously described.[12]

157  Minigene splicing analysis followed the same procedures as for *CFTR* variants, except that two

158 WT constructs (and their corresponding variant constructs) were generated. The *BAP1* Exon 9 insert

159 comprised 337 bp from intron 8, exon 9, and 302 bp from intron 9, whereas the *BAP1* Exon 8–10

160 insert comprised 197 bp from intron 7, exon 8, intron 8, exon 9, intron 9, exon 10, and 243 bp from

161 intron 10.

NM_004656.4 and NG_031859.1 were used as the mRNA and genomic reference sequences of

BAP1, respectively.

Results

Analytical approach

Building on our 2019 findings that some GT>GC variants retain WT transcript,[12] we analyzed three

successive variant categories (Figure 1). First, we re-examined variants previously characterized by

FLGSA. Second, we considered disease-associated GT>GC variants reported since 2019 that generate

WT transcript. Third, we evaluated all possible GT>GC variants in *CFTR* and functionally tested four

selected variants. Across these categories, we assessed residual WT transcript levels, genetic and

clinical findings, SpliceAI ΔDL scores, and database classifications, where available. In addition, we

conducted minigene splicing analysis of the *BAP1* c.783+2T>C variant, which has conflicting

classifications.[17,18] Integrating these findings, we delineated the classification challenges posed by

variants with intermediate functional effects—hereafter referred to conceptually as grey-zone variants.

Comparing disease-associated and FLGSA-analyzed GT>GC variants

In our previous study,[12] the proportion of GT>GC variants retaining WT transcript was similar

between disease-associated and FLGSA-analyzed variants: 15.6% (7 of 45) versus 18.4% (19 of 103).

The magnitude of residual WT transcript levels, however, differed markedly. The seven disease-

associated variants yielded only low levels, from trace amounts up to 15% normal, whereas of the ten

variants generating exclusively WT transcript (as judged by gel electrophoresis), eight retained levels

ranging from 13.7% to 84% of normal, with only two falling below 5%.

The consistently low WT transcript levels observed for the seven disease-associated variants most

likely reflect clinical ascertainment bias, as noted previously.[12] Nonetheless, of the four variants

retaining 10–15% WT transcript, three—*CAV3* (MIM: 601253) c.114+2T>C (GenBank:

NM_033337.3, NG_008797.2),[25] *CD40LG* (MIM: 300386) c.346+2T>C (GenBank: NM_000074.3,

NG_007280.1),[26] and *DMD* (MIM: 300377) c.8027+2T>C (GenBank: NM_004006.3,

NG_012232.1)[27]—were associated with milder clinical phenotypes; the remaining variant, *SPINK1*

c.194+2T>C (GenBank: NM_001379610.1, NG_008356.2), is discussed below.

The corresponding WT constructs for the 103 FLGSA-analyzed substitutions were invariably

preselected to yield a single or quasi-single RT–PCR band of the expected WT transcript size,

consistent with their respective reference mRNA sequences.[12] We also analyzed *HBB* (MIM: 141900)

c.315+2T>C (GenBank: NM_000518.5, NG_059281.1),[28] despite its WT construct producing two

strong bands—one corresponding to the WT transcript and the other representing an aberrantly spliced

transcript. This variant also generated WT transcript in FLGSA,[12] bringing the total number of WT

transcript–generating GT>GC variants identified by FLGSA to 20.

Detailed analysis of three FLGSA-analyzed GT>GC variants

Of the 20 GT>GC variants previously shown by FLGSA[12] to generate residual WT transcript, only

three were found to be registered in ClinVar (https://www.ncbi.nlm.nih.gov/clinvar/; assessed

September 5, 2025). These three variants, described below, illustrate how the presence of residual WT

transcript can challenge variant classification.

SPINK1 c.194+2T>C

Among GT>GC variants, *SPINK1* c.194+2T>C is distinctive in that transcript expression has been

assessed both *in vivo* (using RNA from an affected homozygote)[29] and *in vitro* using FLGSA.[12] Both

approaches yielded concordant results, with approximately 10% of normal WT transcript levels from

the variant allele.[30] SpliceAI predicts a ΔDL score of 0.33, consistent with a substantial yet incomplete

reduction in normal splicing.

Our in-depth analysis of *SPINK1*-related chronic pancreatitis[8] indicated that this ~10% residual

WT transcript places c.194+2T>C in the grey zone, conferring a well-defined genetic effect that is too

weak for "pathogenic" under current ACMG/AMP criteria[5] but too strong for the proposed "risk

allele" category.[7] We therefore suggested classifying it as "predisposing," reflecting its partial LoF

nature and intermediate genetic impact.[8] ClinVar lists this variant as pathogenic/likely pathogenic

(VCV000132142.62; accessed October 28, 2025), but convergent evidence from functional assays, *in*

silico prediction, and clinical and genetic data supports its role as proof of concept highlighting the

need for expanded classification frameworks.[8]

HBB c.315+2T>C

This variant was identified during routine thalassemia screening in a 34-year-old individual who

exhibited hypochromia and microcytosis consistent with a thalassemia trait rather than β-thalassemia

major. The individual was compound heterozygous for *HBB* c.315+2T>C and *HBB* c.79G>A

(p.Glu27Lys) (GenBank: NM_000518.5).[28] The original authors reported that *HBB* c.315+2T>C had a

limited effect on splicing efficiency, stating that it "results in expression that is approximately two-

thirds of the normal allele."[28] However, this estimate was based on hemoglobin quantitation data and

was likely confounded by the *trans*-inherited *HBB* c.79G>A (p.Glu27Lys) variant. The latter likely

retains substantial normal function, as most individuals homozygous for *HBB* c.79G>A (p.Glu27Lys)

exhibit only mild hypochromic microcytic anemia.[31]

Our FLGSA analysis confirmed that *HBB* c.315+2T>C allows appreciable production of WT

transcript.[12] Here, using ImageJ (https://imagej.net/ij/), we compared the intensity of the WT transcript

band derived from the variant allele with that from the WT allele in the original gel.[12] The variant

allele was estimated to retain approximately 30% of WT transcript output, providing a more direct and

reliable measure of its residual transcript level. Notably, the orthologous rabbit variant also generated

WT transcripts,[32,33] supporting conservation of the partial splicing effect across species. The SpliceAI

ΔDL score for *HBB* c.315+2T>C was 0.72, consistent with a moderate predicted loss of the canonical

donor site.

HBB alleles in β-thalassemia have traditionally been classified as β⁰ (complete loss of β-globin

synthesis), β⁺ (moderate-to-severe reduction), or β⁺⁺ ("silent," minimal reduction).[34] *HBB* c.315+2T>C

was originally described as a mild β⁺ allele,[28] which agrees with our FLGSA finding of partial but not

complete loss of WT transcript.[12] Nevertheless, it is currently annotated in ClinVar as "pathogenic"

(VCV000869355.2; accessed October 28, 2025)—a designation that may warrant re-evaluation in

view of its substantial residual WT transcript generation and the mild hematological phenotype

observed in the 34-year-old compound heterozygote.

MGP c.94+2T>C

MGP (MIM: 154870) c.94+2T>C (GenBank: NM_000900.5, NG_023331.2) was shown by FLGSA[12]

to retain approximately 80% of WT transcript, whereas SpliceAI predicted a strong splice-disrupting

effect (ΔDL score, 0.97). Despite this discrepancy, ClinVar lists *MGP* c.94+2T>C as "likely

pathogenic," based on a single clinical testing submission (SCV003031993.2; accessed October 28,

2025) with unknown affected status and no published case reports. This classification appears to rest

on four implicit assumptions:

(1) the variant alters a canonical donor splice site;

(2) such variants generally disrupt splicing;

(3) disruption of canonical sites typically abolishes protein function; and

(4) LoF variants in *MGP* are known to be pathogenic.

As this example shows, ClinVar classifications may at times rely on inferred reasoning rather than

direct experimental evidence.

Disease-associated GT>GC variants reported since 2019

We next examined disease-associated GT>GC variants reported since 2019, focusing on those with

WT transcript generation documented in samples derived from affected individuals. Among the five

variants identified, *SBDS* (MIM: 607444) c.258+2T>C (GenBank: NM_016038.4, NG_007277.1)[35]

and *SRP68* (MIM: 604858) c.184+2T>C (GenBank: NM_014230.4, NG_137597.1),[36] produced only

trace (~2%) or low amounts of WT transcript, with SpliceAI ΔDL scores of 0.93 and 0.37,

respectively. *TNXB* (MIM: 600985) c.12469+2T>C (GenBank: NM_001365276.2, NG_008337.2)

retained a moderate level of WT transcript by allele-specific RT–PCR,[37] consistent with its very low

ΔDL score of 0.01, and has been associated with moderate Ehlers–Danlos–like manifestations in

congenital adrenal hyperplasia.[37] The remaining two variants—*BRCA2* (MIM: 600185) c.8331+2T>C

(GenBank: NM_000059.4, NG_012772.3) and *DOCK8* (MIM: 611432) c.3234+2T>C (GenBank:

NM_203447.4, NG_017007.1)—are discussed below.

BRCA2 c.8331+2T>C

BRCA2 c.8331+2T>C is associated with a high SpliceAI ΔDL score (0.93) and is listed in ClinVar as

"pathogenic/likely pathogenic" (VCV000267692.71; accessed October 28, 2025). However, allele-

specific RT–PCR in a heterozygous carrier demonstrated approximately 60% of WT transcript from

the variant allele.[38] A validated *BRCA1/2* history-weighting algorithm, which compares cancer

histories with expectations for pathogenic versus benign variants,[39] yielded a benign call.[38] Moreover,

BRCA2 appears relatively tolerant of reduced transcript dosage, and one study suggested that variants

producing more than 5% WT transcript should, in the absence of other supporting evidence, be

considered VUS.[40] While the original report reclassified c.8331+2T>C from "pathogenic" to a

"VUS,"[38] the combined evidence appears to point toward a benign interpretation under current

ACMG/AMP principles.

DOCK8 c.3234+2T>C

Biallelic LoF variants in *DOCK8* cause autosomal recessive hyper-IgE syndrome-2 (MIM: 243700).

DOCK8 c.3234+2T>C was reported in the homozygous state in an infant with severe viral

infections.[41] T cells from this individual showed a predominance of WT transcript, normal DOCK8

protein accumulation, intact cellular function, and no DOCK8 deficiency phenotype. The clinical

presentation was instead explained by a homozygous *IFNAR1* (MIM: 107450) nonsense variant

(c.1156G>T, p.Glu386Ter; GenBank: NM_000629.3).[41] Despite this evidence for near-complete WT

transcript generation, lack of disease relevance, and a modest SpliceAI ΔDL score (0.49), three of the

four ClinVar submissions for *DOCK8* c.3234+2T>C (VCV000573664.15; accessed October 28, 2025)

classified it as "likely pathogenic," while one classified it as "benign."

 BRCA2 c.8331+2T>C and *DOCK8* c.3234+2T>C exemplify how variants predicted—or

assumed—to be fully disruptive can nevertheless yield intermediate or even near-complete levels of

WT transcript, underscoring the need for direct transcript-level evidence for accurate classification.

These two variants are summarized in Table 1 alongside the three FLGSA-analyzed examples

(*SPINK1* c.194+2T>C, *HBB* c.315+2T>C, and *MGP* c.94+2T>C).

Locus-wide analysis of *CFTR* GT>GC variants

Having examined GT>GC variants that retain WT transcript generation across different genes, we next

asked how frequent and systematic this phenomenon might be within a single locus. We selected

CFTR (GenBank: NM_000492.4, NG_016465.4), in which biallelic LoF variants cause cystic fibrosis

(CF; MIM: 219700),[42,43] as a model system because it harbors 26 GT donor splice sites, is among the

most extensively studied disease-related genes, and benefits from expert-curated databases CFTR2[21]

and CFTR-France[22] that provide authoritative variant annotations.

Building upon our previous[12,30] and current findings, we observed that although some GT>GC

variants retaining WT transcript are associated with high SpliceAI ΔDL scores, most have low scores.

We therefore used SpliceAI ΔDL scores as the starting point for this analysis, complemented by cross-

examination of variant classifications in CFTR2 and CFTR-France (Figure 2A). Notably, all 16 *CFTR*

GT>GC variants that we grouped as "pathogenic" had SpliceAI ΔDL scores greater than 0.80. These

included 13 designated as "CF-causing" in CFTR2, one absent from CFTR2 but listed as "likely

pathogenic" in CFTR-France (c.3468+2T>C), and two absent from both CFTR2 and CFTR-France but

classified by us as "pathogenic" based on clinical and genetic data in CFTR1 (c.2908+2T>C and

c.2988+2T>C). Eight *CFTR* GT>GC variants were not registered in these three databases, of which

seven also scored above 0.80 and one had a score of 0.56 (c.3717+2T>C). The remaining two *CFTR*

GT>GC variants—one listed as a VUS in CFTR-France (c.2490+2T>C) and one reported in CFTR1 in

two individuals with suspected CF (c.4242+2T>C)—had SpliceAI ΔDL scores below 0.56 (Figure

2A,B). Table S1 provides full details of all 26 *CFTR* GT>GC variants, including the rationale for our

classifying two as "pathogenic."

Assuming that the three low-scoring variants (c.2490+2T>C, SpliceAI ΔDL score 0.35;

c.3717+2T>C, ΔDL 0.56; and c.4242+2T>C, ΔDL 0.15) generate WT transcript, at least 11.5 % of the

26 *CFTR* GT>GC variants would retain WT transcript, which is broadly consistent with the 15–18%

frequency observed across genes in our previous study.[12]

Among the three *CFTR* variants with low SpliceAI ΔDL scores, only c.4242+2T>C had previously

undergone functional analysis. Using an expression minigene (EMG) system comprising the full

CFTR coding sequence and intronic regions tailored to the variant under study, generation of residual

WT transcript from the c.4242+2T>C allele was confirmed by Sanger sequencing, and mature CFTR

protein was detected in transfected cells by immunoblotting.[23] The only other *CFTR* GT>GC variant

reported to yield WT transcripts is c.3873+2T>C, although this was mentioned only briefly and

without accompanying experimental data.[23] Notably, c.3873+2T>C is classified as "CF-causing" in

both CFTR2 and CFTR-France, consistent with its high SpliceAI ΔDL score (0.98) (Figure 2B).

We tested the three low-scoring variants together with c.3873+2T>C in minigene assays using

both pET01 and pSPL3 backbones (Figure 3; see also Figure S1). Findings for each variant are

detailed below.

c.2490+2T>C (intron 14)

WT constructs in both minigene backbones generated multiple RT–PCR products, including the WT

transcript. Importantly, WT transcript from the variant allele was detected in the pET01 backbone

(Figure 3A), consistent with its modest SpliceAI ΔDL score (0.35). Although this variant is not

classified by CFTR2, it is listed in CFTR-France as a VUS, having been reported in two asymptomatic

homozygotes. In CFTR1, it was described in the heterozygous state in the partner of an individual

affected by congenital bilateral absence of the vas deferens (Figure 2B). Furthermore, ClinVar lists

this variant under "conflicting classifications of pathogenicity" (VCV000370078.23; accessed October

28, 2025), based on six clinical-testing submissions ("pathogenic" × 3, "likely pathogenic" × 2, and

"uncertain significance" × 1). Only one submission specified a known affected status—homozygosity

for c.2490+2T>C, labeled "pathogenic," in an individual with fetal cystic hygroma

(SCV001737053.1). Another submission, designated "likely pathogenic" (SCV001363801.5),

explicitly noted that "no occurrence of c.2490+2T>C in individuals affected with cystic fibrosis and

no experimental evidence demonstrating its impact on protein function have been reported." Taken

together, the clinical, genetic, computational, and experimental evidence indicate that c.2490+2T>C is

unlikely to represent a pathogenic *CFTR* variant in the context of classic CF.

c.3717+2T>C (intron 22)

No WT transcript from the variant allele was detected in either minigene backbone (Figure 3B),

despite its intermediate SpliceAI ΔDL score (0.56). This variant is not listed in CFTR2, CFTR-France,

CFTR1, or ClinVar (Figure 2B).

c.3873+2T>C (intron 23)

No WT transcript from the variant allele was observed in either backbone (Figure 3C); this finding is

consistent with its high SpliceAI ΔDL score (0.98), its designation as "CF-causing" in both CFTR2

and CFTR-France, and its "pathogenic" classification in ClinVar (VCV000053828.18; accessed

October 28, 2025) (Figure 2B). Moreover, consistent with its description in CFTR1, this variant is

among the *CFTR* alleles recurrently identified in individuals with clinically diagnosed CF in

Norway.[44] Taken together, c.3873+2T>C represents a pathogenic *CFTR* variant causing classic CF.

The level of WT transcript generated from the c.3873+2T>C allele, as previously reported,[23] is

presumably minimal.

c.4242+2T>C (intron 26)

No WT transcript from the variant allele was detected in either minigene backbone (Figure 3D), in

contrast to the earlier EMG report of residual WT transcript and protein production[23] and to its very

low SpliceAI ΔDL score (0.15). Notably, this variant is not listed in CFTR2 or CFTR-France but

appears in CFTR1 as present in two individuals with suspected cystic fibrosis (Figure 2B). ClinVar

lists c.4242+2T>C as "pathogenic/likely pathogenic" (VCV000035884.8; accessed October 28, 2025),

based on three submissions ("pathogenic" × 1 and "likely pathogenic" × 2), all from clinical genetic

testing laboratories. The collection method was described as "clinical testing" for two submissions

(SCV002630543.2 and SCV004675881.2), both with unknown affected status, and as "curation" for

the remaining submission (SCV000052191.3). Taken together, this variant may best be classified as a

VUS at present.

In summary, locus-wide analysis of *CFTR* confirms that a subset of GT>GC variants retain WT

transcript. Functional testing of selected *CFTR* GT>GC variants highlights both inter-assay variability

and the inherent limitations of minigene systems in fully recapitulating native splicing.[24]

Challenges of splicing assays: the case of *BAP1* c.783+2T>C

As illustrated by *CFTR* variants, *in vitro* splicing assays may fail to yield clear or consistent results,

complicating variant classification. This challenge is further exemplified by *BAP1* c.783+2T>C.

Because heterozygous germline LoF variants in *BAP1* underlie tumor predisposition syndrome 1

(TPDS1; MIM: 614327) and *BAP1* c.783+2T>C affects the canonical intron 9 donor splice site

(GenBank: NM_004656.4, NG_031859.1), this variant was initially classified in ClinVar as "likely

pathogenic," as noted by Goldberg et al.[17] These authors reclassified it as a VUS based on the

following considerations:

(1) the variant was identified in seven families (six Jewish Iraqi and one non-Jewish) with

nonclassical *BAP1*-related TPDS1 tumor types;

(2) the variant exhibited low penetrance;

(3) the variant had a relatively high allele frequency in Jewish Iraqi individuals (3.6%);

(4) the variant had a low SpliceAI score (0.08); and

(5) no aberrantly spliced transcripts were detected by RT-PCR using RNA from blood cells of

two probands.

A subsequent study[18] identified a methodological limitation: the forward primer used by Goldberg

et al.[17] was located within exon 9, rendering it incapable of detecting exon 9 skipping. Using a primer

in exon 8, Sculco et al.[18] analyzed *BAP1* c.783+2T>C heterozygotes and observed an alternative

splicing pattern consistent with exon 9 skipping. These authors proposed reinstating a "likely

pathogenic" classification while acknowledging that the effect of *BAP1* c.783+2T>C on generating

aberrant transcripts might be incomplete.

To further clarify this issue, we performed additional splicing assays (Figure 4; Figure S2). We

first attempted FLGSA, as the *BAP1* genomic sequence from start to stop codon (<8 kb) falls within

the pcDNA3.1/V5-His-TOPO insert size we routinely used in this system.[12] However, the WT

construct failed to yield clear RT-PCR bands (Figure 4A). Notably, in our previous study only 33 of

119 (28%) WT full-length constructs produced a single or quasi-single band of the expected size after

transfection.[12] We therefore turned to minigene assays using pET01 and pSPL3. WT constructs

spanning exons 8–10 with flanking introns produced multiple RT-PCR bands, none corresponding to

the WT transcript defined by NM_004656.4, nor did the variant constructs (Figure 4B). In contrast,

shorter constructs containing exon 9 with flanking intronic sequences generated the expected WT band

from the WT allele, whereas only transcripts lacking exon 9 were produced from the variant allele

(Figure 4C; Figure S2).

ClinVar currently lists c.783+2T>C under "conflicting classifications of pathogenicity"

(VCV000422670.29; accessed October 28, 2025), with five submissions of "uncertain significance"

and one of "likely benign." In light of these inconclusive findings, allele-specific transcript analyses in

heterozygous individuals—if possible in the presence of a *cis*-linked coding variant outside exon 9—

would be most informative. Alternatively, targeted knock-in cell models could provide additional

insight into the splicing effect of *BAP1* c.783+2T>C.

Discussion

Our analysis yielded several insights into the impact of GT>GC 5′ splice-site variants. Consistent with

findings from FLGSA,[12] we show that disease-associated GT>GC variants can generate a wide range

of residual WT transcript levels, from barely detectable to nearly normal. Moreover, our findings

confirm our earlier postulate that "GT>GC variants in human disease genes may not invariably be

pathogenic,"[12] as exemplified by the *DOCK8* c.3234+2T>C variant.

Second, GT>GC variants that generate substantial yet incomplete WT transcript levels often fall

within a functional grey zone, complicating their classification (Figure 5). Such indeterminacy

underscores the limitations of binary frameworks and has important clinical implications. A clear

example is *SPINK1* c.194+2T>C, which retains approximately 10 % of normal WT transcript. In the

simple heterozygous state, its effect is too weak to be classified as "pathogenic" yet too strong to be

regarded as a mere "risk" allele. However, in homozygous, compound heterozygous, or *trans*-

heterozygous states—or in the presence of environmental modifiers—the same variant confers a

substantially greater pathogenic effect.[8] This exemplifies how variants with intermediate functional

effects contribute to the genetic complexity of human disease. To address this challenge, we have

proposed expanding the ACMG/AMP variant classification framework by incorporating additional

categories, including a "predisposing" category.[6,8]

Third, our findings underscore the limited resolution of current *in silico* tools, such as SpliceAI,[19]

for predicting whether GT>GC variants generate WT transcript and, if so, to what extent. Whereas

most WT transcript–retaining variants exhibit low SpliceAI ΔDL scores, some have high values

(>0.80), and transcript levels show no consistent correlation with predicted scores. This limitation

partly reflects the intrinsic ability of GC to function as a natural donor site in approximately 1% of U2-

type introns.[45] More broadly, it highlights the difficulty of predicting splicing outcomes that are not

all-or-none, leaving many variants within a functional grey zone that current algorithms cannot readily

model or quantify.

Fourth, our results highlight the methodological challenges inherent in experimentally

characterizing splice-altering variants. RNA from disease-relevant tissues provides the most reliable

readout, yet such material is rarely available, and surrogate sources such as blood cells or

lymphoblastoid lines are informative only if the gene of interest is expressed. In the absence of RNA

from affected individuals, cell-based assays—FLGSA,[12] EMG,[23] midigene,[46] or minigene[47] systems—

are often used, but each has intrinsic limitations. For example, not only is FLGSA impractical for large

genes, but also not all successfully constructed full-length WT constructs yield the expected WT

transcript. The most frequently used minigene assays are performed in a highly artificial background

that lacks the natural genomic context required for proper splicing regulation.[48,49] Such loss of context

is particularly problematic for variants with intermediate effects, for which classification depends not

only on detecting aberrant transcripts but also on quantifying residual WT transcript from the variant

allele. Even when RNA from affected individuals is available, interpretation in heterozygotes can be

confounded by WT transcript derived from the unaffected allele. For this reason, in both our

previous[12] and current studies, analyses of disease-associated GT>GC variants were focused primarily

on homozygotes and hemizygotes. RNA-seq, although increasingly applied in clinical contexts,[50-52]

faces similar challenges of tissue accessibility and allele-specific quantification. Collectively, these

methodological obstacles underscore the need for refined experimental systems, such as those based

on transactivation or transdifferentiation.[53]

GT>GC variants that retain residual WT transcript reflect the degeneracy and context dependence

of splice-site recognition. As such, findings from this study can likely be extrapolated to many other

splice-altering variants—excluding those at the invariant +1, −1, and −2 positions—because *cis*-acting

motifs are degenerate, and deep intronic variants can generate novel splice sites that compete with

physiological ones. Although a comprehensive survey is beyond the scope of this study, several

observations support this view. In SpliceVarDB, approximately 50% of variants could not be assigned

unambiguously to "splice-altering" or "non–splice-altering" categories but instead fell into a "low-

frequency splice-altering" group, corresponding to weak or indeterminate evidence of

spliceogenicity,[54] suggesting that grey-zone variants are common. Among 12 FLGSA-analyzed

SPINK1 coding variants showing aberrant splicing, nine (including presumed missense and

synonymous variants) also generated residual WT transcript.[55] Likewise, deep intronic variants such as

CFTR c.1584+18672A>G (GenBank: NM_000492.4, NG_016465.4)[56] and *CLRN1* c.254−643G>T

(GenBank: NM_174878.3, NG_009168.1)[57] have been reported to yield both aberrant and WT

transcripts. Collectively, these examples indicate that residual WT transcript generation is a recurrent

and underappreciated phenomenon across genes and variant types, producing outcomes along a

continuum rather than at binary extremes (Figure 5). Because residual WT transcript cannot always be

readily assessed in simple or compound heterozygotes, and splice-altering variants account for an

estimated 10–30% of all disease-associated alleles,[11,54,58] we infer that variants generating residual WT

transcript may represent a significant source of variable penetrance and expressivity.

Nonetheless, the boundaries of the functional grey zone (Figure 5) likely vary across genes,

depending on functional tolerance thresholds and modes of inheritance, such that an identical level of

residual WT transcript may be clinically inconsequential in one gene–disease context yet functionally

significant in another. Although developed here in the context of LoF variants for conceptual

simplicity, the grey-zone model may also extend to gain-of-function or dominant-negative variants,[59]

in which disease-associated alleles within a given gene are likewise unlikely to exert uniform effects.

Notably, this conceptual framework was established primarily for genes in which complete or near-

complete LoF alleles cause classical Mendelian diseases. As we have previously opined,[8] some

disease-associated genes may not necessarily harbor clearly "pathogenic" variants.

Taken together, our study establishes GT>GC variants as a tractable model for addressing the

classification challenges posed by splice-altering variants with intermediate functional consequences.

By using residual WT transcript as a measurable parameter—and integrating clinical and genetic data,

RNA analyses of material derived from affected individuals, full-length and minigene assays, and

computational predictions—we show that residual WT transcript generation is a recurrent and

biologically relevant phenomenon that frequently complicates classification. Gradual functional

effects are also observed for other variant types, as exemplified by missense variants in *BRCA1* (MIM:

113705)[60] and *SOD1* (MIM: 147450),[61] both analyzed using multiplexed assays of variant effect

(MAVE). These observations underscore the need to move beyond binary pathogenic/benign

frameworks toward models that accommodate a continuum of effects,[6,8] as efforts focused solely on

refining ACMG/AMP criteria within a binary framework (as reflected in recent studies[62-65]) are

unlikely to resolve the fundamental problem amplified by the field's transition from a "phenotype-

first" to a "genotype-first" approach.

To address this challenge, new conceptual, computational, and experimental strategies are

emerging. As noted by Whitcomb,[66] our recently proposed "pathogenic–predisposing–risk–benign"

framework[8] offers a practical means of addressing the variant classification challenge. The use of

machine-learning models trained on large-scale population-based datasets,[9] together with functional

data from MAVE,[67,68] holds promise for delineating the full spectrum of variant effects revealed

through analyses of increasingly large cohorts.[69-71] Ultimately, grey-zone variants may constitute a

substantial fraction of those currently classified as VUS or filtered out for exceeding the allele-

frequency threshold set for high-penetrance disease variants and could account for part of the "missing

heritability" in both rare and complex diseases.[72-74] Recognizing the broad prevalence of grey-zone

variants and their contribution to genetic complexity will be an important step toward realizing the full

potential of precision medicine.[75]

Our study has several limitations. First, the number of GT>GC 5′ splice-site variants known to

generate substantial amounts of WT transcript remains limited. Second, the level of WT transcript

generated from certain GT>GC variants, as determined by FLGSA or RT-PCR using material derived

from affected individuals, may differ from that occurring under true *in vivo* conditions within the

pathologically relevant tissue. Third, a potential alternative impact of GT>GC variants—namely, a

reduction in overall splicing efficiency without producing aberrant transcripts—could not be assessed

in this analysis. Likewise, GT>GC variants that generate both WT and aberrant transcripts, but in

which the aberrant transcripts are degraded by nonsense-mediated decay (NMD),[76] could not be

evaluated. These two scenarios could lead to underestimation of the true extent of partial LoF GT>GC

variants. Fourth, while we highlighted the classification challenges posed by the variants discussed in

detail, we did not systematically attempt to specify how each variant should be classified. This is

primarily because, as we have previously noted, "defining precise thresholds between different variant

categories remains challenging and must be tailored to each gene and disease context."[8] Despite these

limitations, our integrative analysis of GT>GC variants provides a coherent framework for

understanding how residual WT transcript generation shapes the functional continuum of splice-

altering variants and informs their broader biological and clinical interpretation.

In conclusion, viewing GT>GC variants through the lens of residual WT transcript generation

offers valuable insight into how partial LoF splice-altering variants—and, more broadly, variants with

intermediate functional effects—can contribute to genetic complexity, complicate both *in silico*

prediction and experimental analysis, and challenge variant classification (Figure 5).

Data and code availability

537   All data supporting this study are provided within the article and its supplementary materials. This

538   study did not generate any new datasets or code.

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Author contributions

Conceptualization, J.M.C.; study design, J.H.L., H.W., W.B.Z. and J.M.C.; methodology, J.H.L., H.W., W.B.Z. and J.M.C.; experiments, J.H.L., H.W., X.Y.T. and W.B.Z.; literature search and variant collation, E.M. and J.M.C.; writing—original draft preparation, J.M.C.; writing—review and editing, J.H.L., H.W., X.Y.T., E.M., D.N.C., C.F., Z.L., W.B.Z. and J.M.C.; project administration, W.B.Z. and J.M.C.; funding acquisition, J.H.L. and J.M.C. All authors have read and approved the final version of the manuscript and agree to be accountable for all aspects of the work, ensuring that questions related to the accuracy or integrity of any part of the work are appropriately investigated and resolved.

Declaration of interests

The authors declare no competing interests.

Declaration of generative AI and AI-assisted technologies in the writing process

During the preparation of this work the authors used ChatGPT 5 in order to improve readability and language. After using this tool, the authors reviewed and edited the content as needed and take full responsibility for the content of the publication.

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Figure 1. Scheme of the analytical approach

GT>GC variants were analyzed in three successive categories, together with minigene analysis of a *BAP1* variant with conflicting classifications. Within each category, variants were examined for four features (yellow shading). The red box highlights the central issue addressed in this study—the classification challenges posed by grey-zone variants that retain substantial yet incomplete wild-type transcript generation.

A

All 26 possible GT>GC substitutions in *CFTR*, categorized by SpliceAI ΔDL scores and database information

>0.80

Grouped as "pathogenic" (n = 16)

Not in CFTR2, CFTR1, or CFTR-France (n = 7)

≤0.56

Not in CFTR2, CFTR1, or CFTR-France (n = 1)

Remaining variants (n = 2)

Minigene splicing assay

B

CFTR variants selected for functional analysis using minigene splicing assays

Variant	SpliceAI ΔDL score	Database classification and reporting				Residual wild-type transcript	
		CFTR2	CFTR-France	CFTR1	ClinVar	Previous study	This study
c.2490+2T>C	0.35	–	VUS (reported in two asymptomatic homozygotes)	Identified in heterozygous state in the partner of a CBAVD patient	CCP	–	Yes (pET01); No (pSPL3)
c.3717+2T>C	0.56	–	–	–	–	–	No in both assays
c.3873+2T>C	0.98	CF-causing	CF-causing	Commonly reported in Norwegian CF patients	P	Yes	No in both assays
c.4242+2T>C	0.15	–	–	Reported in two individuals with suspected CF diagnosis	P/LP	Yes	No in both assays

Figure 2. Analysis of GT>GC substitutions in *CFTR*

(A) Flowchart summarizing the 26 possible *CFTR* GT>GC substitutions categorized by SpliceAI ΔDL scores and database information (see Table S1 for details). Variants with low scores (≤0.56) were selected for minigene splicing assays.

(B) Four variants analyzed by minigene splicing assays. In "Database classification and reporting," note that, unlike CFTR2 and CFTR-France, CFTR1 does not classify registered variants. In "Residual wild-type transcript," "Yes" and "No" indicate whether the variant allele generated wild-type transcript. Results from this study are shown separately for the two assay systems (pET01 and pSPL3; see Figure 3), with "No in both assays" indicating absence of wild-type transcript from the variant allele in both systems. "Previous study" refers to Joynt et al.[23]

Abbreviations: CBAVD, congenital bilateral absence of the vas deferens; CF, cystic fibrosis; CCP, conflicting classifications of pathogenicity; ΔDL, delta donor loss; P, pathogenic; P/LP, pathogenic/likely pathogenic; VUS, variant of uncertain significance.

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Figure 3. Minigene splicing analysis of four selected *CFTR* GT>GC substitutions in pET01 and

pSPL3 backbones

Each variant was analyzed with its preceding exon and short flanking intronic regions cloned into the

respective reporter vectors. Arrows indicate bands corresponding to the normally spliced transcript, in

accordance with the *CFTR* mRNA reference sequence NM_000492.4. Additional bands are annotated

in Figure S1.

Figure 4. Splicing analysis of the *BAP1* c.783+2T>C variant in different assay contexts

(A) Full-length gene splicing assay (FLGSA). The wild-type (WT) construct failed to generate clear reverse transcription–PCR (RT–PCR) bands.

(B) Minigene analysis of exons 8–10 with flanking intronic sequences in pET01 and pSPL3 backbones. No normally spliced transcripts were detected, even from the WT constructs. For band annotations, see Figure S2.

(C) Minigene analysis of exon 9 with flanking intronic sequences in pET01 and pSPL3 backbones. Arrows indicate bands corresponding to the normally spliced transcript. Additional bands are annotated in Figure S2.

GenBank *BAP1* mRNA and genomic reference sequences: NM_004656.4 and NG_031859.1.

Figure 5. Conceptual model of the functional grey zone in variant classification

Residual wild-type (WT) transcript levels and variant classification lie along continua rather than binary states. Variants with intermediate WT transcript levels often fall within a functional grey zone, where *in silico* prediction and experimental analysis become increasingly difficult. Such grey-zone variants may act through variant–variant (both intra- and intergenic) or variant–environment interactions, thereby contributing to the genetic complexity of human disease. Although illustrated here with splice-altering variants, this conceptual model also applies to other variant types and disease mechanisms shaped by graded levels of functional impact.

888   **Table 1. Five GT>GC variants with residual WT transcript that present classification challenges**

Variant[a]	Reference mRNA and genomic sequences	hg38 coordinate	Residual WT transcript[b]	Assessment method	SpliceAI ΔDL score	Clinical/genetic findings	ClinVar classification
SPINK1 c.194+2T>C	NM_001379610.1, NG_008356.2	chr5-147828020-A-G	10%	RT–PCR of material from an affected homozygote;[29] FLGSA[12]	0.33	Well-defined genetic effect, too weak for pathogenic classification[8]	Pathogenic/likely pathogenic
HBB c.315+2T>C	NM_000518.5, NG_059281.1	chr11-5226575-A-G	30%	FLGSA[12]	0.72	Originally described as a mild β^+-thalassemia allele[28]	Pathogenic
MGP c.94+2T>C	NM_000900.5, NG_023331.2	chr12-14884211-A-G	80%	FLGSA[12]	0.97	Not reported in affected individuals	Likely pathogenic
BRCA2 c.8331+2T>C	NM_000059.4, NG_012772.3	chr13-32363535-T-C	60%	Allele-specific RT–PCR of affected individual–derived material[38]	0.93	*BRCA2* history-weighting algorithm[39] indicated a benign profile;[38] *BRCA2* exhibits tolerance to reduced transcript dosage[40]	Pathogenic/likely pathogenic
DOCK8 c.3234+2T>C	NM_203447.4, NG_017007.1	chr9-399261-T-C	Predominant WT	RT–PCR and protein analyses of T cells from an affected homozygote[41]	0.49	Disease explained by an unlinked homozygous *IFNAR1* nonsense variant[41]	Conflicting classifications of pathogenicity

889   Abbreviations: ΔDL, delta donor loss; FLGSA, full-length gene splicing assay; RT-PCR, reverse transcription-PCR; WT, wild-type.

890   [a]Variants are arranged in order of mention in the main text.

891   [b]Percentage of WT transcript generated from the variant allele relative to that from the WT allele.