

Article

Not peer-reviewed version

Exploring Novel Applications: Repositioning Clinically Approved Therapies for Medulloblastoma Treatment

[Arthur Karaulic](#) and [Gilles Pagès](#) *

Posted Date: 21 July 2025

doi: 10.20944/preprints2025071627.v1

Keywords: Medulloblastoma; kinase inhibitors; immune check point inhibitors; mutations; non-metastatic/metastatic tumors; BCL2; Venetoclax



Preprints.org is a free multidisciplinary platform providing preprint service that is dedicated to making early versions of research outputs permanently available and citable. Preprints posted at Preprints.org appear in Web of Science, Crossref, Google Scholar, Scilit, Europe PMC.

Copyright: This open access article is published under a Creative Commons CC BY 4.0 license, which permit the free download, distribution, and reuse, provided that the author and preprint are cited in any reuse.

Disclaimer/Publisher's Note: The statements, opinions, and data contained in all publications are solely those of the individual author(s) and contributor(s) and not of MDPI and/or the editor(s). MDPI and/or the editor(s) disclaim responsibility for any injury to people or property resulting from any ideas, methods, instructions, or products referred to in the content.

Article

Exploring Novel Applications: Repositioning Clinically Approved Therapies for Medulloblastoma Treatment

Arthur Karaulic and Gilles Pagès *

University Côte d'Azur, Institute for Research on Cancer and Ageing of Nice (IRCAN^o,
UMR CNRS 7284/U INSERM 1081

* Correspondence: gpages@unice.fr; Tel.: +33-4-89153479

Key Points

- Overexpression of targets of clinically approved treatments correlates with either longer or shorter survival.
- A treatment designed for hematological tumors, venetoclax, appear relevant for medulloblastoma.
- The presence of mutations may warrant treatment efficacy as personalized medicine.

Abstract

Background/Objectives: The advent of kinase inhibitors (TKI), therapeutic antibodies and inducers of apoptosis has revolutionized cancer treatment, yet their application in pediatric tumors, particularly medulloblastoma, remains understudied. Understanding the expression of these targets in specific genetic subgroups could unveil potential repositioning opportunities for already approved drugs. **Methods:** We analyzed RNA-sequencing data from the R2 Genomics Analysis and Visualization Platform (N = 763 patients, multiple cohorts) and the TCGA database (six individual cohorts 828 patients) to assess the expression of 73 potential targets of TKIs and antibodies targeting immune checkpoint inhibitors (ICI) or membrane receptors and inducers of apoptosis. These treatments, FDA-approved or in phase II clinical trials for solid or hematologic cancers, and their targets were evaluated in both non-metastatic and metastatic patients when data was available. Additionally, we examined treatments tailored to mutated targets crucial for tumorigenesis or resistance to conventional therapies. **Results:** Overexpression of certain targets beyond predefined cutoff values in Kaplan-Meier analyses correlated with either prolonged or shortened overall survival. Targets associated with shorter survival suggested potentially relevant treatments, thereby highlighting the importance of defining specific treatments for distinct genetic subgroups. Notably, certain immune checkpoint inhibitors showed relevance for specific subgroups but detriment for others. As a positive control, our analysis confirmed the use of axitinib, an anti-angiogenic treatment, as demonstrated by our recent publication. Surprisingly, a treatment developed for hematological tumors, venetoclax, demonstrated potential efficacy in medulloblastoma. **Conclusions:** Medulloblastoma displays subtype-specific expressions of FDA-approved TKI, ICI and pro-apoptotic drug targets, impacting overall survival. Clinical trials investigating these approved treatments in medulloblastoma are therefore warranted.

Keywords: Medulloblastoma; kinase inhibitors; immune check point inhibitors; mutations; non-metastatic/metastatic tumors; BCL2; Venetoclax

1. Introduction

Medulloblastoma patients undergo highly intensive treatments involving surgery, radiotherapy (for patients above 3 years old), and intensive multimodal chemotherapies. However, such aggressive

treatments pose significant risks for young patients during critical growth periods. Consequently, there is a pressing need to explore strategies for de-escalating treatment intensities without compromising efficacy or exacerbating side effects. Targeted, less toxic therapies are imperative to enhance cure rates, manage relapses, and mitigate the adverse effects associated with current intensive regimens.

Over the past 15 years, considerable advancements have been made in cancer treatments across various types. Yet, the challenge persists in developing innovative therapies, often constrained by logistical hurdles and financial constraints. Herein lies an opportunity to reposition existing treatments through a molecular pathology approach, leveraging the analysis of specific mutation expressions in newly diagnosed patients or those experiencing relapses. By harnessing available data from public databases, we can tailor treatments to specific genetic subgroups of medulloblastoma, thus creating a customized treatment landscape.

This inclusive genetic approach allows us to construct a comprehensive performance profile for existing treatments. Surprisingly, treatments approved for hematological cancers show promise in medulloblastoma, suggesting potential avenues for repurposing therapies. By adopting this streamlined methodology, we uncover compelling opportunities for treatments to undergo evaluation in early-phase clinical trials, paving the way for meaningful advancements in medulloblastoma management.

Medulloblastoma ranks as the second most prevalent solid pediatric tumor. Pediatric medulloblastoma is composed of several molecular subgroups: Wingless (WNT), Sonic Hedgehog (SHH), Group 3 and Group 4. Groups WNT and SHH are characterized by abnormal activation of the corresponding signal transduction. Group 3 and Group 4 show overexpression of N- and c-Myc, inactivation of p53 and deleterious chromosomal abnormalities [1]. Patients undergo rigorous treatments comprising surgery, radiotherapy (administered to patients above 3 years old), and chemotherapy regimens incorporating agents like etoposide, carboplatin, and vincristine. While these intensive interventions yield a 70% long-term remission rate, most patients suffer from the severe side effects of these treatments [2,3] and approximately 30% of cases experience relapse, with fatal outcomes being commonplace. Therefore, physicians face two paramount challenges in managing medulloblastoma: 1) **Reducing Treatment Intensity**: The primary objective is to mitigate the high detrimental side effects associated with intensive therapies, which can encompass mobility issues, cognitive impairments, language deficits, and motor function limitations. This reduction must be achieved without compromising treatment efficacy; 2) **Identifying Relevant Therapies for Relapse**: Another critical goal is to propose effective treatments in instances of relapse, with the aim of extending survival and potentially achieving a second long-term remission. Addressing these challenges demands a delicate balance between treatment effectiveness and the minimization of adverse effects, underscoring the importance of tailored therapeutic strategies in the management of medulloblastoma.

Despite advancements in treatment strategies, the development of new therapies for pediatric cancers remains challenging, largely due to historical underinvestment by the pharmaceutical industry. Treatment repositioning represents a promising alternative, leveraging the efficacy of several targeted therapies that have been successfully utilized for decades in adult cancers sharing similar mechanisms of tumor aggressiveness with medulloblastomas. For instance, reverse genetic analysis conducted by Coy et al. [4] demonstrated the potential application of antibody-drug conjugates for pediatric brain tumors.

Given the hyper vascularized nature of medulloblastomas and considering that increased angiogenesis is associated with the most aggressive medulloblastomas [5]. We, along with German collaborators, have highlighted the utility of Axitinib, a tyrosine kinase inhibitor initially approved for metastatic kidney cancer [6] or combined with immune checkpoint inhibitors. Axitinib has shown promise both as a monotherapy and in combination with immune checkpoint inhibitors for kidney cancers [7] but also for treating medulloblastomas [8,9]. This approach has led to the initiation of the clinical trial **Mependax** (NCT06485908), which is currently enrolling patients with relapsed

medulloblastoma and ependymoma, building upon encouraging findings in six children treated with Axitinib and metronomic etoposide [10].

To identify additional therapeutic candidates, we implemented a systematic approach. First, we cataloged available targeted therapies, including older and next-generation agents, alongside their molecular targets. Utilizing publicly accessible datasets via the R2 platform https://hgserver1.amc.nl/cgi-bin/r2/main.cgi?open_page=login and the TCGA database through cBioportal <https://www.cbioportal.org/>, we analyzed the expression of target genes and their correlation with survival outcomes. Specific activating mutations in target genes were also evaluated, given that many therapies are indicated only for tumors harboring such mutations.

This strategy identified several unexpected candidates for repositioning in medulloblastoma, including therapies traditionally used for hematological malignancies. The validity of our approach was supported by the identification of Axitinib [8] and HER2 inhibitors [4,11], as relevant treatment options. Furthermore, we demonstrated the potential of Venetoclax, typically used in hematological tumors, as a promising therapeutic candidate for medulloblastoma.

2. Materials and Methods

2.1. Expression and Mutation Profiling

Gene expression levels and survival correlations using the Kaplan Meier method were analyzed using the R2 Genomics Analysis and Visualization Platform (https://hgserver1.amc.nl/cgi-bin/r2/main.cgi?open_page=login) and the cBioportal platform (<https://www.cbioportal.org/>). For the R2 platform, medulloblastoma datasets included the following:

- GSE85217 (Cavalli, [12])
- GSE67851 (Hsieh, TH [13])
- GSE37418 (Gilbertson, RJ [14])
- GSE74195 (den Boer, M [15])
- GSE10327 (Kool, M [16])
- GSE49243 (Pfister 2, [17])
- GSE12992 (Delattre, [18])
- GSE3526 (Roth, [19])
- Cohort Pfister 1 [20]

For the cBioportal platform, the following data set were analyzed:

- Medulloblastoma (PCGP, Nature 2012, [14])
- Medulloblastoma (Broad, Nature 2012, [21])
- Medulloblastoma (DKFZ, Nature 2017, [20])
- Medulloblastoma (ICGC, Nature 2012, [22])
- Medulloblastoma (Sickkids, Nature 2016, [23])

These datasets provided robust platforms for exploring gene expression profiles and their association with clinical outcomes.

2.2. Establishment of the List of Targeted Therapies

The compilation of targeted therapies, including kinase inhibitors, immune system modulators, and specific monoclonal antibodies, was carried out using the resources provided by the National College of Medical Pharmacology (<https://pharmacomedicale.org>) (Table 1). This approach ensured a comprehensive and up-to-date selection of therapeutic agents relevant to the study's focus.

2.3. Cell Lines

The human medulloblastoma cell lines (ONS76, DAOY, HD-MB03) and the normal microglial cells (HMC3) were purchased from American Type Culture Collection (ATCC). DAOY, ONS76 and HMC3 cells were maintained in MEM alpha (Gibco, Life Technologies Corporation, Loughborough, UK) supplemented with 10% fetal bovine serum (FBS, SIGMA, Burlington, MA, USA). HD-MB03

cells were maintained with RPMI 10% fetal bovine serum (FBS, SIGMA, Burlington, MA, USA). Cells were monitored routinely, and the absence of mycoplasma was verified monthly using the Plasmotest kit (Invivogen, San Diego, CA, USA).

2.4. Cell Death Assay

Cell viability was assessed using the propidium iodide (PI) exclusion assay. Following treatment, cells were harvested and incubated with PI (10 µg/ml) for 5 minutes. The percentage of PI-positive cells was subsequently analyzed by flow cytometry using a MACSQuant Analyzer (Miltenyi Biotec, catalog number 130-092).

2.5. Immunoblot

Cells were lysed with Laemmli buffer and protein amounts were determined by the Pierce TM BCA Protein Assay Kit (Thermo Fisher). Then, 20 µg of protein were resolved by SDS-PAGE. The proteins were transferred onto PVDF membranes in Tris-glycine buffer. Membranes were blocked with 5% milk at room temperature and then immunoblotted overnight in 3% milk with the anti-BCL2 antibody (Cell signaling Technology #2872). Membranes were washed with PBS-Tween 0.1% and incubated with HRP-conjugated secondary antibodies at room temperature for 1 h. The Advanta Western Bright Quantum HRP substrate was used as a detection reagent.

3. Results

3.1. Identifying Medulloblastoma Patients Eligible for Targeted Therapies

Table 1 highlights kinase and apoptosis inhibitors that target specific genes in tumor cells and are used in the treatment of both hematological and solid tumors. It also includes immunological modulators that act on specific targets expressed by transformed immune cells or on pathways involved in immune tolerance. Our aim was to explore the potential of repositioning these clinically validated therapies for the treatment of pediatric medulloblastoma.

To validate their relevance, we analyzed the relationship between these genes and overall patient survival (OS) using the Cavalli et al. cohort, the only publicly available dataset with survival data [12]. For each gene associated with a targeted therapy, Kaplan-Meier survival curves were generated via the R2 platform, employing the optimal cutoff values. Initially, Kaplan-Meier survival curves were generated for each gene, accompanied by their respective raw P-values and Bonferroni-corrected P-values. Approximately 300 curves were produced, which are presented in **Supplementary Figure S1**. Based on this analysis, **Table 2** summarizes the raw and Bonferroni-corrected P-values for each gene across the molecular subgroups of medulloblastoma (WNT, SHH, Group 3, and Group 4).

From **Table 2**, several genes were identified as relevant therapeutic targets due to their overexpression being associated with shorter OS. The corresponding treatments are outlined in **Table 3**. Genes with statistically significant associations (raw and Bonferroni-corrected P-values < 0.05) were prioritized for potential efficacy of targeted therapy within specific genetic subgroups:

SHH subgroup: Relevant targets included BCR-ABL, FILP1, PDGFR, Kit, CD52, FGFR, RET, RAF, TIM3, and VEGFR, with corresponding treatments Imatinib, Alemtuzumab, Pemigatinib, Regorafenib, Sabatolimab, Axitinib and Lenvatinib.

Group 3 tumors: Key genes identified were BCL2 and MEK2, with potential treatments Venetoclax and MEK inhibitors such as Cobimetinib, Trametinib, or Selumetinib.

WNT subgroup: Targetable genes included ABL1, CDK4/6, FGFR, HER2, and SMO, with suggested therapies such as Asciminib, Abemaciclib, Palbociclib, Ribociclib, Pemigatinib, Trastuzumab, Pertuzumab, T-DM1, T-DXd, Lapatinib, Tucatinib, and Sonidegib.

Table 1. Overview of Approved Targeted Therapies for Solid and Hematologic Tumors. This table provides a detailed summary of targeted therapies currently approved for the treatment of solid and hematologic tumors. It includes the associated pathologies, the generic names of the treatments, their corresponding brand names, and the specific genes they target.

	Name of "INIB" and "UMAB"	Brand name	Target gene(s)
Immunomodulation	Baricitinib	OLUMIANT	JAK1, JAK2
	Tofacitinib	XELJANZ	JAK1, JAK3
Hematology	Ruxolitinib	JAKAVI	JAK (+JAKV617F)
	Acalabrutinib	IMBRUVICA	BTK (Bruton)
	Idelalisib	ZYDELIG	P13K
	Venetoclax	VENCLYXTO	BCL2
	Midostaurin	RYDAPT	FLT3, KIT, VEGFR, PKC
	Ivosidenib	TIBSOVO	IDH1
	Gilteritinib	XOSPATA	FLT3, AXL
	Asciminib	SCEMBLIX	ABL1, BCR-ABL1
	Bosutinib	BOSULIF	BCR-ABL, SRC, LYN, HCK, PDGFR
	Dasatinib	SPRYCEL	BCR-ABL, SRC-KINASES
	Ponatinib	ICLUSIG	KIT, FLT3, RET, PDGFR, VEGFR
Nilotinib	TASIGNA	BCR-ABL	
Imatinib	GLIVEC	BCR-ABL, FILP1, PDGFR, KIT	
Solid Tumors	Avapritinib	AYYAKYT	PDGFR, KIT
	Sunitinib	SUTENT	VEGFR, PDGFR, KIT, FLT3
	Pazopanib	VOTRIENT	VEGFR, PDGFR, KIT, FLT3
	Axitinib	INLYTA	VEGFR, PDGFR, KIT, FLT3
	Cabozantinib	CABOMETYX	VEGFR, MET, RET
	Sorafenib	NEXAVAR	VEGFR, PDGFR, KIT, FLT3, RAF-KINASES
	Lenvatinib	LENVIMA	VEGFR, FGFR, RET, PDGFR, KIT
	Vandetanib	CAPRELSA	VEGFR, EGFR, RET
	Pemigatinib	PEMAZYRE	FGFR
	Regorafenib	STIVARGA	KIT, RET, RAF-KINASES
	Soridegib	ERIVEDGE	SMO (Hedgehog pathway)
	Lorlatinib	LORVIQUA	ALK, ROS1
	Brigatinib	ALUNBRIG	ALK, ROS1
	Ceritinib	ZYKADIA	ALK
	Crizotinib	XALKORI	HGFR, c-MET
	Alectinib	ALECENSA	ALK, RET
	Afatinib	GIOTRIF	HER1, 2, 3, 4
	Dacomitinib	VIZIMPRO	HER1, 2, 4, DDR2
	Erlotinib	TARCEVA	HER1
	Gefitinib	IRESSA	HER1
	Osimertinib	TAGRISSO	EGFR T790M
	Dabrafenib	TAFINLAR	BRAF V600E
	Encorafenib	BRAFTOVI	BRAF V600E
	Vemurafenib	ZELBORAF	BRAF V600E
	Cobimetinib	COTELLIC	MEK
	Trametinib	MEKINIST	MEK
	Selumetinib	KOSELUGO	MEK
	Capmatinib	TRABECTA	RET fusion
	Selpercatinib	RET SEVMO	RET fusion
	Praseltinib	GAVRETO	RET fusion
	Larotrectinib	VITRAKVI	TRK
	Alpelisib	PIQRAY	P13Ka
	Lapatinib	TYVERB	HER2
	Tucatinib	TUKYSA	HER2
Abemaciclib	VERZENIOS	CDK4/6	
Palbociclib	IBRANCE	CDK4/6	
Ribociclib	KISQALI	CDK4/6	
Olaparib	LYNPARZA	PARP	
Rucaparib	RUBRACA	PARP	
Talazoparib	TALZENNA	PARP	
Immune cell targeting	Blinatumomab	BLINCYTO	CD19
	Basiliximab	SIMULECT	CD25
	Gemtuzumab ozogamicin	MYLOTARG	CD33
	Alemtuzumab	CAMPATH	CD52
	Sabatolimab		TIM3
	Ipilimumab	YERVOY	CTLA4
	Adecatumumab		EPCAM
Atezolizumab, avelumab, durvakumab	TECENTRIO, BAVENCIO, IMFINZI	PDL1	

Group 4 tumors: Although no targets were highly statistically significant, several potential candidates were identified, including ABL1 (Asciminib), BCR-ABL, FILP1, PDGFR, Kit (Imatinib), BCL2 (Venetoclax), CD33 (Gemtuzumab), CD52 (Alemtuzumab), CDK4/6 (Abemaciclib, Palbociclib, Ribociclib), CTLA4 (Ipilimumab), JAK1/3 (Tofacitinib), MEK (Cobimetinib, Trametinib, Selumetinib), PARP (Olaparib, Rucaparib, Talazoparib), PDL1 (Atezolizumab, Avelumab, Durvalumab), PDGFR/KIT (Avapritinib), and TIM3 (Sabatolimab).

This classification provides a framework for repositioning specific drugs to treat pediatric medulloblastoma, offering novel therapeutic opportunities tailored to the genetic profiles of the tumor subgroups.

Table 2. Association Between Targeted Genes and Patient Survival Across Medulloblastoma Subgroups. This table illustrates the relationship between genes targeted by the therapies listed in Table 1 and overall survival (OS) across various genetic subgroups of medulloblastoma. The analyzed genes or gene families are displayed alongside their respective P-values, calculated using the R2 platform with the optimal cutoff method. Two P-values are provided for each gene: the first represents the raw significance, while the second reflects the Bonferroni-corrected significance. Genes are visually categorized based on their prognostic association with background colors: **White background:** Genes linked to shorter OS; **Black background:** Genes linked to longer OS; **Dark grey background with enlarged text:** Genes associated with a poor prognosis, indicated by both raw and Bonferroni-corrected P-values < 0.05; **Light grey background:** Non-significant (NS) genes with no clear survival impact. This classification provides an intuitive visual summary of the survival impact of specific genes across medulloblastoma subgroups, facilitating a better understanding of their prognostic relevance.

	ALK	BCL2	BCL2L1	BCL2L2	BTX	CD3D	CD19	CD20	CD25	CD33	CD52	CDK	DDR2	EPCAM	FGFR	FGFR3	FLT3	
WNT	NS	NS	NS	NS	3e-3NS	NS	NS	0.02NS	NS	NS	NS	NS	NS	NS	NS	NS	0.04NS	
SHH	1e-3NS	0.03NS	NS	NS	0.01NS	2e-2NS	0.04NS	1.4e-3NS	0.01NS	2e-3NS	2.8e-40.04	9e-3NS	0.07NS	1.4e-3NS	NS	NS	9e-50.01	0.04NS
GR4	0.02NS	0.02NS	1e-3NS	NS	NS	9e-40.02	4.2e-4NS	2e-40.07	NS	0.01NS	1.5e-4NS	0.01NS	2e-3NS	3e-3NS	NS	NS	5e-3NS	7e-3NS
GR3	1.5e-3NS	5e-40.05	NS	NS	3e-3NS	0.04NS	0.03NS	0.05NS	NS	0.01NS	6.7e-4NS	NS	2e-40.02	0.03NS	0.04NS	NS	NS	0.05NS

	HER				Immune check points				IDH1	JAK	MAPK						
WNT	EGFR/HER1	ERBB2/HER2	ERBB3/HER3	ERBB4/HER4	PD1	PDL1	PDL2	CTLA4	TIGIT	LAG3	IDH1	JAK1	JAK2	JAK3	MAPK2	MAPK1	MAPK3
SHH	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS
GR4	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS
GR3	0.02NS	0.01NS	NS	0.02NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS

	MET	MTOR	NTRK	PARP1	PDGFR	PIK3G
WNT	NS	NS	NS	NS	NS	NS
SHH	0.03NS	NS	1e-7/1.5e-5	0.01NS	5e-3NS	7e-50.01
GR4	0.03NS	NS	0.07NS	NS	0.02NS	3e-3NS
GR3	0.02NS	0.01NS	0.02NS	NS	0.01NS	0.02NS

	PKC		RAF		RET	ROS1	SMO	SRC			VEGFR							
WNT	ITPKC	PRKCA	PRKCB	PRKCD	ARAF	BRAF	RAF1	RET	ROS1	SMO	ABL1	SRC	FYN	YES	LYN	FLT1	KDR	FLT4
SHH	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS
GR4	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS
GR3	0.02NS	0.04NS	0.02NS	NS	0.01NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS	NS

Table 3. Alignment of Targeted Therapies with Gene-Survival Associations Across Medulloblastoma Subgroups. This table outlines the positioning of targeted therapies based on the survival impact of their corresponding genes across medulloblastoma subgroups. It highlights the most suitable treatments for each subgroup, determined by the lowest P-value: **Small characters:** Indicate a trend toward significance. **Medium characters:** Denote a statistically significant raw P-value. **Large characters:** Represent both raw and Bonferroni-corrected P-values as statistically significant. This table aids in identifying the most effective targeted therapies for each medulloblastoma subgroup, guided by gene-survival correlations.

TARGET	WNT	SHH	GR4	GR3
ABL1	Asciminib	Asciminib	Asciminib	
BCR-ABL, FILP1, PDGFR, KIT		Imatinib	Imatinib	
BCL2		Venetoclax	Venetoclax	Venetoclax
BTK		Acalabrutinib		Acalabrutinib
CD19/CD3		Blinatumomab		
CD25		Basiliximab		
CD33		Gemtuzumab	Gemtuzumab	
CD52		Alemtuzumab	Alemtuzumab	Alemtuzumab
CDK4/6	Abemaciclib, Palbociclib, Ribociclib	Abemaciclib, Palbociclib, Ribociclib	Abemaciclib, Palbociclib, Ribociclib	
CTLA4		Ipilimumab	Ipilimumab	Ipilimumab
EPCAM				Adecatumumab
FGFR	Femigatinib	Pemigatinib		Gilteritinib
FLT3				
HER2	Trastuzumab, Pertuzumab, TDM1, TDXI, Lapatinib, Tucatinib			Trastuzumab, Pertuzumab, TDM1, TDXI, Lapatinib, Tucatinib
JAK1, 3			Tofacitinib	
KIT RET RAF		Regorafenib		
MEK		Cobimetinib, Trametinib, Selumetinib	Cobimetinib, Trametinib, Selumetinib	Cobimetinib, Trametinib, Selumetinib
MET				Crizotinib
PARP		Olaparib, Rucaparib, Talazoparib	Olaparib, Rucaparib, Talazoparib	Olaparib, Rucaparib, Talazoparib
PDL1			Atezolizumab, Avelumab, Durvalumab	
PDGFR, KIT			Avapritinib	
PIK3CA		Idelalisib		Idelalisib
SMO	Sonidegib	Sonidegib		
TIM3		Sabatolimab	Sabatolimab	Sabatolimab
VEGFR, EGFR, RET		Vandetanib		
VEGFR, FGFR, RET, PDGFR, KIT		Axitinib, Lenvatinib		



in M0 Group 4 patients but poor prognosis in M1 Group 4 patients. Similarly, **CTLA4** was a favorable prognostic marker in M0 Group 3 patients but indicated poor prognosis in M1 Group 3 patients.

These results highlight that the prognostic behavior of certain genes depends significantly on tumor stage, reinforcing the importance of metastatic status in defining their clinical relevance. However, most genes demonstrate consistent prognostic behavior across both stages.

Genes associated with good prognosis were identified as being more significant in different patient groups based on metastatic status. In M0 patients, the genes **BTK, CD3D, CD19, CD20, CD33, DDR2, FGFR1, HER1, HER2, HER3, HER4, PD1, PDL1, PDL2, TIM3, LAG3, IDH1, JAK1, JAK2, MAPK1, NTRK1, NTRK3, PIK3CA, PIK3CD, PIK3C2A, PIK3C2G, PRKCB, PRKCD, RAF1, ROS1, LYN, KDR, and FLT4** demonstrated greater significance. Meanwhile, in M1 patients, genes such as **CD19, CD25, CDK6, FGFR3, PD1, MET, MTOR, NTRK2, PDGFRB, FIP1L1, PIK3C3, BRAF, ROS1, FYN, LYN, and FLT4** were identified as more significant.

Conversely, genes associated with poor prognosis also showed variation in significance based on metastatic status. In M0 patients, the genes **BCL2, BTK, CD3D, CD25, CD33, CD52, CDK4, DDR2, EPCAM, FGFR2, FGFR3, HER3, PD1, PD2, PDL2, TIM3, JAK3, MAP2K2, MAPK1, NTRK1, PARP1, PDGFRA, PDGFRB, FIP1L1, KIT, PIK3G, PIK3C2B, PRKCA, PRKCD, BRAF, RAF1, RET, SMO, ABL1, SRC, FYN, and KDR** were more significant. In contrast, in M1 patients, the genes **BTK, CD3D, CD19, CDK4, CDK6, DDR2, FLT3, HER1, HER4, PD2, PDL1, JAK1, MAPK3, NTRK1, NTRK2, PDGFRA, KIT, PIK3G, PIK3C2A, PIK3C2B, PIK3C2G, ITPKC, PRKCD, ARAF, RAF1, ROS1, FYN, FLT1, and KDR** showed higher significance.

These findings underscore the critical role of metastatic status in determining the prognostic significance and therapeutic prioritization of target genes. This distinction is vital for guiding gene-specific treatment strategies and optimizing therapeutic outcomes in medulloblastoma.

Through the deconvolution of genes associated with aggressiveness, we identified targeted therapies tailored to each genetic subgroup and their corresponding metastatic status (**Table 5**).

This updated table incorporates considerations for both genetic subgroups and tumor stages, in contrast to the previous table, which focused solely on genetic subgroups without accounting for tumor metastatic status. For the WNT subgroup, several therapies showed promise for M0 patients, who represented most of the cohort. These included Abemaciclib, Ribociclib, and Palbociclib (CDK inhibitors), Pemigaptinib (FGFR inhibitor), and, unexpectedly, Sonidegib (SMO inhibitor), which was originally designed for SHH patients. However, targeting HER2 was not beneficial for M0 patients in this subgroup.

In the SHH subgroup, Venetoclax (**BCL2 inhibitor**), Ipilimumab (**CTLA4 checkpoint inhibitor**), and Alpelisib (**PIK3CA inhibitor**) were found to be irrelevant for both M0 and M1 patients. Conversely, therapies such as Gilteritinib (**FLT3 inhibitor**), Sunitinib, Pazopanib, and Axitinib (**targeting VEGFR, PDGFR, KIT, and FLT3**) demonstrated potential greater effectiveness.

For Group 4, therapies such as Asciminib (**ABL1 inhibitor**), Gemtuzumab (**CD33-directed agent**), Ipilimumab (**CTLA4 inhibitor**), and Cobimetinib, Trametinib, and Selumetinib (**MAP2K2 inhibitors**), and Sabatolimab (**TIM3**), were no longer relevant for either M0 or M1 patients. However, potential treatments included Gilteritinib (**FLT3 inhibitor**), Gefitinib, Erlotinib, Afatinib, and Cetuximab (**EGFR-targeting agents**), as well as Regorafenib (**targeting KIT, RET, and RAF**), Nivolumab and Pembrolizumab (**PD1 inhibitors**), and Idelalisib (**PIK3CA inhibitor**).

For Group 3, Sabatolimab (**TIM3**) was suspected to have no effect for both M0 and M1 patients. In contrast, promising drugs included Abemaciclib, Ribociclib, and Palbociclib (**CDK inhibitors**), Gefitinib, Erlotinib, Dacomitinib, and Cetuximab (EGFR-targeting), along with Ivosidenib (**IDH1 inhibitor**), Regorafenib (**targeting KIT, RET, and RAF**), and Cabozantinib (**targeting VEGFR, MET, and RET**). These findings underscore the importance of considering differences in tumor stages—localized (M0) versus metastatic (M1)—when repositioning treatments. The choice of therapy should be informed by these distinctions, particularly at diagnosis, relapse, or based on whether the tumor is localized or has spread.

Table 5. Alignment of Targeted Therapies with Gene-Survival Associations Across Medulloblastoma Subgroups Considering M0 and M1 Status. This table aligns targeted therapies with the survival impact of their corresponding genes across medulloblastoma subgroups, considering the metastatic status (M0: non-metastatic; M1: metastatic). The therapies are categorized based on the statistical significance of the association between the targeted gene and overall survival (OS), as determined by the lowest P-value. The significance levels are represented using text size: Small characters: Indicate a trend toward significance (suggestive but not statistically confirmed); Medium characters: Denote a statistically significant association based on raw P-values; Large characters: Represent statistically significant associations confirmed by both raw and Bonferroni-corrected P-values. This framework identifies the most promising therapies for each subgroup, offering a nuanced understanding of therapeutic relevance in the context of metastatic status and gene-survival dynamics.

TARGET	WNT M0	WNT M1	SHH M0	SHH M1	GR4 M0	GR4 M1	GR3 M0	GR3 M1
ABL1			Aciclovir, Nilotinib					
BCR-ABL, FILP1, PDGFR, KIT			Imatinib		Imatinib	Imatinib		
BCL2					Venetoclax	Venetoclax	Venetoclax	Venetoclax
BTK			Acabrutinib					Acabrutinib
CD19/CD3			Blinatumomab	Blinatumomab				
CD25			Basilimab					
CD33			Gemtuzumab	Gemtuzumab			Gemtuzumab	
CD52			Alemtuzumab	Alemtuzumab	Alemtuzumab	Alemtuzumab	Alemtuzumab	
CDK4/6	Abemaciclib, Ribociclib, Palbociclib		Abemaciclib, Ribociclib, Palbociclib		Abemaciclib, Ribociclib, Palbociclib	Abemaciclib, Ribociclib, Palbociclib	Abemaciclib, Ribociclib, Palbociclib	Abemaciclib, Ribociclib, Palbociclib
CTLA4								Ipilimumab
EPCAM							Adecatumumab	
FGFR	Pemigatinib		Pemigatinib		Pemigatinib	Pemigatinib		
FLT3				Gilteritinib	Gilteritinib			Gilteritinib
HER1			Gefitinib, Erlotinib, Afatinib, Cetuximab		Gefitinib, Erlotinib, Cetuximab	Gefitinib, Erlotinib, Afatinib, Cetuximab	Gefitinib, Erlotinib, Dacomitinib, Cetuximab	
HER2							Trastuzumab, Pertuzumab, Lapatinib, Ado-trastuzumab	
IDH1								ivosidenib
JAK1, 2, 3					Tofacitinib	Tofacitinib, Baricitinib		
KIT, RET, RAF					Regorafenib	Regorafenib	Regorafenib	
MEK			Cobimetinib, Trametinib, Selumetinib				Cobimetinib, Trametinib, Selumetinib	
MET							Capotecinib, Crizotinib	
PARP			Olaparib, Rucaparib, Talazoparib		Olaparib, Rucaparib, Talazoparib		Olaparib, Rucaparib, Talazoparib	Olaparib, Rucaparib, Talazoparib
PD1					Nivolumab, Pembrolizumab			
PDL1					Atezolizumab, Avelumab, Durvalumab	Atezolizumab, Avelumab, Durvalumab		
PDGFR, KIT					Avapritinib	Avapritinib		
PIK3CA							Idecalixib	Idecalixib
SMO	Sonidegib		Sonidegib		Sonidegib			
TIM3			Sabatimab	Sabatimab				
VEGFR, EGFR, RET			Vandetanib					
VEGFR, FGFR, RET, PDGFR, KIT			Lenvatinib					
VEGFR, PDGFR, KIT, FLT3			Sunitinib, Pazopanib, Axitinib	Sunitinib, Pazopanib, Axitinib				

3.3. Analysis of Targetable Mutations in Medulloblastoma Cohorts

Table 1 highlights treatments applicable when specific mutations are present, such as **BRAFV600E**. To explore actionable targets, we analyzed medulloblastoma cohorts from the TCGA database, focusing on mutations in targetable genes. Across six available cohorts comprising 828 patients, we identified specific mutation frequencies as follows: 54 patients (6.5%) carried mutations in **CTNNB1**, 42 patients (5.1%) in **PTCH1**, 21 patients (2.5%) in **SMO**, 14 patients (1.7%) in **SUFU**, 13 patients (1.6%) in **PTEN**, 5 patients (0.6%) in **PIK3CA**, 3 patients (0.36%) in **PIK3R1**, and 2 patients (0.24%) in **FGFR1**. Additionally, mutations were identified in each single patient (0.12%) for **ATM**, **CDKN2A**, **ERBB4**, **FGFR2**, **IDH1**, and **NRAS**.

Given these mutation profiles, conventional treatments listed in **Table 1** may not be effective. The specific mutations identified in these cohorts and their clinical implications are detailed in **Table 6**.

To address this, we propose alternative therapies tailored to mutation-specific cases. For mutations in the **CTNNB1** pathway, inhibitors of GSK3 such as Elraglusib and Tideglusib [26,27] which target CTNNB1 activation, may be effective. In the **PI3K/AKT** pathway, several inhibitors could be utilized, including Capivasertib, used for metastatic breast cancer with at least one alteration on PIK3CA/AKT1/PTEN [28], and RLY-2608, a PIK3CA inhibitor [29]. Additionally, PI3K β inhibitors such as GSK2636771 and AZD8186 are potential options [30,31]. For mutations in the **FGFR** pathway, inhibitors like Erdafitinib and Fexagratinib may offer therapeutic [32,33]. In cases with EGFR mutations that heterodimerize with ERBB4, the inhibitor Dacomitinib could be considered [34].

Mutations associated with resistance to SHH pathway inhibitors (e.g., Sonidegib and Vismodegib) were also identified, underscoring their role as predictive markers of treatment inefficacy.

Table 6. Alignment of Targeted Therapies with Specific Gene Mutations This table presents the alignment of targeted therapies with specific mutations that are not addressable by conventional therapies. It includes the names of patient cohorts from the TCGA, the specific genes and mutations identified, their associated expected phenotypes, and the names of treatments tailored to these mutations.

Cohort Name	Mutation (Number of case)	Phenotype	Treatment
PCGP Medulloblastoma Nature 2012 37 patients CTNNB1 (beta Cat) 6 Patients 5% PIK3CA 1 patient 2.7 % PTCH1 1 patient 2.7 %	Mutations S33F (1), G34R (1) Mutation Q648K (1) Mutation S881Mfs*2 (1)	Likely Oncogenic Oncogenic Likely oncogenic truncating mutation	Eraglusib, Tideglusib Capivasertib Sonidegig Vismodegib
Medulloblastoma Broad Nature 2012 92 Samples CTNNB1 (beta Cat) 6 Patients 6.5 % PTCH1 5 Patients 6.5 % FGFR1 2 Patients 2.2 % SMO 1 Patient 1.1 % PTEN 1 Patient 1.1 % PIK3CA 1 Patient 1.1 % SUFU 1 Patient 1.1 %	Mutation D32G (1), Mutation S33F (1), S33Y (1) G34R (2), S37P (1) Mutation X400 Splice (1), I1055Sfs*3 (1), Q242Vfs*9 (1) Q889Afs*7 (1), n V439Qfs*59 (1) Mutation N577K (1), K687E (1) Mutation L412F (1) Mutation G165E (1) Mutation H1047L (1) Mutation X388 Splice (1)	Likely Oncogenic Likely oncogenic truncating mutations Likely oncogenic truncating mutation Likely Oncogenic Likely Oncogenic Oncogenic Oncogenic Likely Oncogenic	Sonidegig Vismodegib Sonidegig Vismodegib Erdafinib, Fezagranib Resistance to Vismodegib Capivasertib, GSK2636771, AZD8186 Apelcisb Capivasertib RLY-2008 Resistance to Vismodegib
Medulloblastoma DKFZ Nature 2017 491 Samples PTCH1 25 Patients 5.1 % CTNNB1 (beta Cat) 25 Patients 5.1 % ERBB4 1 Patient 0.2 % SMO1 11 Patients 2.2 % SUFU 11 Patients 2.2 % IDH1 1 Patient 0.2 % PTEN 9 Patients 1.83 % PIK3R1 3 Patients 0.61 %	Mutation C92* (1), W129* (1), Q177* (1), E374* (1), F583* (1) R602* (1), C818* (1), Q839* (1), C1093* (1), E1183* (1) Mutation V169Pfs*92 (1), Q242Vfs*9 (1), C226Pfs*2 (1) Mutation D301fs*23 (1), X400_splice (1), L448Nfs*9 (1) L450Pfs*5 (1), A451Pfs*5 (1), C462Wfs*27 (1) F465Nfs*50 (1), X750_splice (1), Y797Sfs*24 (1) L819Tfs*10 (1), I1055Sfs*3 (1), X1150_splice (1) Mutation D32Y (2), S33C (2), S33F (6), S33Y (4), G34E (1) G34R (4), G34V (1), S37C (1), S37F (1), S37P (1) S37Y (1), T41A (1) Fusion ERBB4-LCLAT1 (1) Mutation L412F (8), W535L (3) Mutation I291* (1), E437* (1), P12Rfs*90 (1), A25Gfs*23 (1) Y147Vfs*21 (1), F149Sfs*20 (1), T190Nfs*25, T261Gfs*8 (1) P341Rfs*20 (1), X388_splice (1), SUFU-CYP17A1-AS1 Fusion (1) Mutation R132C Mutation W111R (1), G132V (1), G165E (1), D324Y (1), M134del (1) T319fs*2 (1), K322Rfs*23 (1), PTEN-NR2F1-AS1 Fusion (1) Mutation PTEN-THAP9 Fusion (1) X582_splice (1), Y452_N453ins* (1), D68Gfs*38 (1)	Likely oncogenic truncating mutations Likely Oncogenic Likely Oncogenic Likely Oncogenic Oncogenic Oncogenic Likely Oncogenic Oncogenic Likely Oncogenic	Sonidegig Vismodegib Eraglusib, Tideglusib Lapatinib Dacomitinib Resistance to Vismodegib Resistance to Vismodegib Ivodesinib Vorasidenib Capivasertib GSK2636771 AZD8186 Capivasertib
Medulloblastoma ICGC Nature 2012 125 Samples CTNNB1 (beta Cat) 15 Patients 12 % PTCH1 6 Patients 4.8 % PIK3CA 2 Patients 1.6 % NRAS 1 Patient 0.8 % CDKN2A 1 Patient 0.8 % SMO 8 Patient 6.4 % PTEN 1 Patient 0.8 % PIK3R1 1 Patient 0.8 % ATM 1 Patient 0.8 % FGFR2 1 Patient 0.8 %	Mutation D32A (2), D32Y (3), S33C (4), S33F (2) G34R (2), S37F (1), S37Y (1) Mutation E374* (1), Y1009* (1), F465Efs*50 (1) D773Rfs*16 (1), Y804Cfs*3 (1), L819Tfs*10 (1) Mutation H1047L (1), C420R (1) Mutation G13V Mutation D94N Mutation L412F Mutation H03Y Mutation X582_splice Mutation Y1915* Mutation K659E	Likely Oncogenic Likely oncogenic truncating mutations Oncogenic Oncogenic Likely oncogenic truncating mutation Likely Oncogenic Oncogenic Likely Oncogenic Likely Oncogenic Likely Oncogenic	Eraglusib, Tideglusib Sonidegig Vismodegib Apelcisb Capivasertib RLY-2008 Binimetinib Cobimetinib Trametinib Palbociclib Ribociclib Abemaciclib Resistance to Vismodegib Capivasertib GSK2636771 AZD8186 Capivasertib Olaparib Talazoparib + Enzalutamide Erdafinib RLY-4008 AZD4547
Medulloblastoma PCGP Nature 2012 37 Samples CTNNB1 (beta Cat) 4 Patients 10.8 % PIK3CA 1 Patient 2.7 % SUFU 2 Patients 5.4 %	Mutation D32G (1), S33F (1), S33Y (1), G34R (1) Mutation Q546K Mutation T261Gfs*8 (1), V148Sfs*30 (1)	Likely Oncogenic Oncogenic Likely Oncogenic	Eraglusib, Tideglusib Capivasertib + Fulvestrant RLY-2008 Resistance to Vismodegib
Medulloblastoma Sickkids Nature 2016 46 Samples PTCH1 4 Patients 8.7 % PTEN 2 Patients 4.3 % CTNNB1 (beta Cat) 2 Patients 4.3 % SMO 1 Patient 2.2 %	Mutation E675* (1), V442Gfs*54 (1), S444As*11 (1), Y452Lfs*4 (1) Mutation A126T (1), G132V (1) Mutation D32V (1), G34R (1) Mutation L412F	Likely oncogenic truncating mutation Oncogenic Likely Oncogenic Likely Oncogenic	Sonidegig Vismodegib Capivasertib GSK2636771 AZD8186 Eraglusib, Tideglusib Resistance to Vismodegib

This analysis provides valuable insights into potential repurposing of drugs approved for other tumors to address specific subsets of medulloblastoma, enhancing treatment precision.

3.4. Experimental Validation of In Silico-Identified Therapies in Medulloblastoma Cell Lines

Based on Table 5, several unexpected treatments emerged as potentially relevant for medulloblastoma, including Venetoclax, a BCL2-targeted therapy traditionally used for hematological malignancies and seldom applied to solid tumors. [35]. Our analysis indicated that this therapy showed the greatest efficacy in Group 3 and Group 4 tumors, with a lesser effect on SHH

tumors. To validate this, we compared BCL2 expression in normal cerebellum and different datasets of medulloblastoma in the R2 database. This analysis reveals a global upregulation of BCL2 across several datasets of medulloblastoma compared to healthy cerebellum (Figure 1A). Then, we assessed BCL2 protein levels in normal microglial cells (HMC3) and in the medulloblastoma cell lines ONS76 (SHH), DAOY (SHH with a P53 mutation mimicking Group 3 tumor outcomes), and HDMB03 (Group 3 tumor). BCL2 expression, as assessed by immunoblot analysis, was highest in DAOY cells, intermediate in ONS76 and HMC3 cells, and low in HDMB03 cells (Figure 1B). Then, we compared the effect of Venetoclax in normal cells based on their neural origin and tumor cells. The IC50 of Venetoclax was determined on these normal cells and medulloblastoma cell lines (Figure 1 C-F). Except in normal cells, dose-response experiments demonstrated a direct correlation between BCL2 expression levels and Venetoclax efficacy (Figure 1C-G). subsequently calculated the specificity index for venetoclax (Figure 1G) [36]. The specificity index was defined as the ratio of the IC50 in HMC3 (normal) cells and the IC50 in medulloblastoma cell lines, with values greater than 1 indicating preferential activity in tumor cells. Although a specificity index above 5 is generally considered indicative of therapeutic relevance with low toxicity [36], Venetoclax has already demonstrated both efficacy and manageable toxicity in a pediatric cohort of patients with newly diagnosed acute myeloid leukemia [37]. These clinical findings support Venetoclax as a viable treatment option in children. Taken together with our data, these results suggest that Venetoclax warrants further preclinical evaluation as a potential therapy for medulloblastoma.

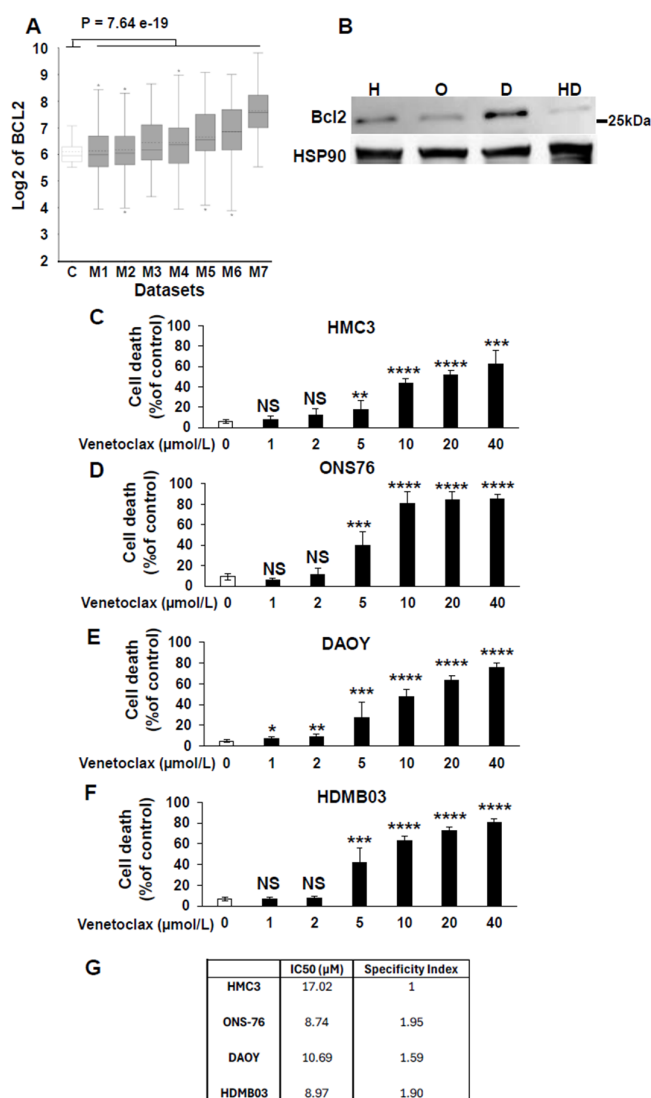


Figure 1. Relevance of targeting BCL2 for the treatment of medulloblastoma. A) Box-dot plot showing mRNA expression level of BCL2 (R2: Genomics Analysis and Visualization Platform (<http://r2.amc.nl>) in normal cerebellum (C) (n = 9 retrieved from Roth database), and medulloblastoma patients (M1–M7); M1: (n = 76) from Gilbertson database, M2: (n = 57) from Delattre database, M3: (n = 62) from Kool database, M4: (n = 223) from Pfister1 database, M5: (n = 51) from den Boer database, M6: (n = 31) from Hsieh database and M7: (n = 73) from Pfister2 database. Statistical analysis (One-way analysis of variance (ANOVA)) is shown, $P = 7.64 \times 10^{-19}$. B) Immunoblot analysis of BCL2 expression in HMC3 (H) ONS76 cells (O), DAOY cells (D) and HDMB03 cells (HD). GAPDH is shown as loading control. C-F) Dose dependent inhibition of viability of the four cell lines by increasing concentrations of venetoclax. *: $P < 0.05$; **: $P < 0.01$; ***: $p < 0.001$; **** $P < 0.0001$. G) IC_{50} and specificity index of venetoclax for the different cell lines.

4. Discussion

The treatment of medulloblastoma remains a significant challenge, whether in the first line setting— aiming to limit the side effects of intensive therapies—or following relapse, which often represents a therapeutic dead end. The development of new drugs is particularly complex and time-intensive, often taking years to reach clinical application. In this context, drug repositioning offers a promising alternative, bypassing extensive preclinical toxicology studies in animals, though still requiring clinical trials to assess toxic effects in children, who are no longer viewed as “small adults.”

Through comprehensive analysis of available patient cohorts, we identified potential opportunities for therapy repositioning using drugs with established safety profiles in adults. **Table 7** highlights several targeted therapies previously tested in early-phase clinical trials for pediatric brain tumors. Some demonstrated promising effects, though confirmation in phase III trials is still required. Notably, therapies relevant to specific genetic subgroups of patients in our analysis were often found ineffective when metastatic status was considered, underscoring a potential source of failure in some clinical trials. Additionally, the presence of mutations in specific genes may act as confounding factors in therapeutic outcomes.

Among the repositioned therapies, treatments for hematological cancers such as **Imatinib** and **Venetoclax** emerged as particularly intriguing. **Venetoclax** showed significant specificity in medulloblastoma model cell lines, further supporting its potential relevance considering that BCL2 expression correlated to the most aggressive form of medulloblastoma [38]. Moreover, using a completely different approach, Garancher, A. et al. demonstrated the efficacy of the pan-inhibitor of BCL family members, TW37 [39]. However, based on our study, the use of a pan-BCL family inhibitor is not recommended, as the overexpression of certain BCL family members appears to have a beneficial effect for reasons that remain unclear. Instead, **Venetoclax** is strongly recommended due to its selective targeting and its ability to penetrate the blood-brain barrier [40].

Table 7. Overview of treatments that have been approved for clinical use or previously described in the literature. Potentially repositionable treatments were extensively reviewed to evaluate their relevance for medulloblastoma or other brain tumors. The table also indicates whether each treatment can cross the blood-brain barrier (BBB) and provides the corresponding PubMed ID (PMID) for reference.

TREATMENT	Clinical application				Research (in vitro/vivo)				Cross the BBB	Reference(s)
	Used in MB	Reference(s)	Other brain tumor	Reference(s)	Used in MB	Reference(s)	Other brain tumor	Reference(s)		
Abemaciclib	No		No		No		Glioblastoma	PMID: 3859583	Yes (low)	PMID: 3859583
Acalabrutinib	No		Leptomeningeal/CNS	PMID: 35732356	No		Glioblastoma	PMID: 34577576		
Afatinib	No		No		No		Neuroblastoma (-in vivo)	PMID: 32705581		
Alentuzumab	No		CD4+ / T cells in brain	PMID: 3632454	No		No		Yes	PMID: 3632454
Alpelisib	No		Toxile / CNS	PMID: 32802528	Yes (+in vivo)	PMID: 31492956	No			PMID: 33728134
Alpelisib	No		No		Yes (+in vivo)	PMID: 31492956	Neuroblastoma (-in vivo)	PMID: 33491755		
Asciminib	No		No		Yes (+in vivo)	PMID: 37781987	Glioblastoma	PMID: 33318517		
Atezolizumab	No		Glioblastoma	PMID: 30073642	No		No		Yes	PMID: 37122727
Avapritinib	No		No		No		No			PMID: 38167404
Avetumab	No		No		No		No			
Axitinib	Yes	PMID: 38778441	No		No		No		Yes	PMID: 35008234
Axitinib	Yes	PMID: 38778441	No		PMID: 35008234		No			PMID: 31035676
Axitinib	Yes	PMID: 38778441	No		PMID: 34234256		No			
Axitinib	Yes	PMID: 38778441	No		PMID: 29377550		No			
Axitinib	Yes	PMID: 38778441	No		PMID: 31035676		No			
Baricitinib	No		No		No		No			
Basiliximab	No		No		No		No			
Blinatumomab	No		No		No		No			
Cetuximab	Yes	PMID: 23426003	No		No		No			
Cetuximab	Yes	PMID: 24952425	No		No		No			
Cobimetinib	No		No		No		No			
Dacomitinib	No		No		Yes (+in vivo)	PMID: 29574250	Glioblastoma & Pilocytic astrocytoma (-in vivo)	PMID: 29574250		
Durvalumab	No		No		non		No			
Erlotinib	Yes	PMID: 29778738	Glioblastoma	PMID: 32550606	Yes	PMID: 21726539	Glioblastoma	PMID: 21726539		
Gefitinib	No		No		Yes (+in vivo)	PMID: 19033425	Glioma	PMID: 18829483		
Gefitinib	No		No		Yes	PMID: 18829483				
Gemtuzumab	No		No		No		No			
Gilteritinib	No		No		No		No			
Idelalisib	No		No		No		No			
Imatinib	No		No		Yes	PMID: 37387713	Glioblastoma	PMID: 37387713		
Imatinib	No		No		PMID: 13417143					
Ipilimumab	Yes	PMID: 36808285	Glioblastoma	PMID: 36808285	Yes	PMID: 31533380	No			
Nilotinib	No		No		Yes	PMID: 31533380	No			
Nivolumab	Yes	PMID: 36808285	Glioblastoma	PMID: 36808285						
Nivolumab	Yes	PMID: 36808285	Glioblastoma	PMID: 36808285						
Olaparib	No	PMID: 36681550	No		Yes	PMID: 37783879	neuroblastoma (-in vivo)	PMID: 34508175		
Olaparib	No	PMID: 36681550	No		Yes (+in vivo)	PMID: 35001340	Glioblastoma	PMID: 35001340		
Olaparib	No	PMID: 36681550	No		Yes (+in vivo)	PMID: 34508175	High grade glioma and Ependymoma	PMID: 22184287		
Olaparib	No	PMID: 36681550	No		Yes	PMID: 22184287				
Olaparib	No	PMID: 36681550	No		Yes (+in vivo)	PMID: 35005517	No			
Olaparib	No	PMID: 36681550	No		Yes (+in vivo)	PMID: 37333134				
Olaparib	No	PMID: 36681550	No		Yes	PMID: 37127632				
Olaparib	No	PMID: 36681550	No		Yes (+in vivo)	PMID: 35080286				
Palbociclib										
Palbociclib										
Palbociclib										
Palbociclib										
Pembrolizumab	non		No		No		No			
Pemigatinib	non		No		No		No			
Pertuzumab	non		No		No		No			
Ribociclib	non		No		Bioinformatic study	PMID: 37783879	No			
Ribociclib	non		No		Yes (+in vivo)	PMID: 36318650				
Ribociclib	non		No		Yes (+in vivo)	PMID: 35709750				
Ribociclib	non		No		Yes	PMID: 20971695				
Rucaparib	No		No		No		No			
Sabatolimab	No		No		No		No			
Selumetinib	No		No		Yes (+in vivo)	PMID: 35835837	No			
Selumetinib	No		No		Yes (+in vivo)	PMID: 29301011				
Sonidegib										
Sunitinib	No		No		Yes (+in vivo)	PMID: 35008234	No		Yes	PMID: 35008234
Sunitinib	No		No		Yes	PMID: 26524040				PMID: 26524040
Sunitinib	No		No		Yes	PMID: 20553726				
Talazoparib	No		No		Yes (+in vivo)	PMID: 37919621	No		No	PMID: 37919621
Talazoparib	No		No		Yes (+in vivo)	PMID: 25263539				
Tofacitinib	No		No		Yes	PMID: 30332965	No			
Trametinib	No		No		Yes	PMID: 37726268	No			
Trastuzumab	No		No		No		No			
Venetoclax	No		No		Yes	PMID: 35061150	No			
Venetoclax	No		No		Yes (+in vivo)	PMID: 33855528				
Venetoclax	No		No		Yes	PMID: 26264044				

5. Conclusions

Although our findings represent an initial step in treatment repositioning for pediatric brain tumors, they provide a foundation for exploring therapies previously overlooked for medulloblastoma, with the goal of improving outcomes in this challenging context.

Supplementary Materials: The following supporting information can be downloaded at the website of this paper posted on Preprints.org, Figure S1: Kaplan-Meier analysis of the relationship between targetable genes and survival in different medulloblastoma subgroups; Figure S2: Kaplan-Meier analysis of the relationship between targetable genes and survival in different medulloblastoma subgroups considering the non-metastatic (M0) and metastatic (M1) patients.

Author Contributions: Conceptualization, G.P. methodology, G.P.; formal analysis, G.P.; investigation, G.P. and A.K.; resources, G.P.; data curation, G.P. and A.K.; writing—original draft preparation, G.P.; writing—review and editing, G.P. and A.K.; supervision, G.P.; project administration, G.P.; funding acquisition, G.P. All authors have read and agreed to the published version of the manuscript.

Funding: This research was funded by W the Conseil Général 06, the FEDER, the EUR HEALTHY from the University Cote d'Azur, the Région Provence-Alpes-Côte d'Azur, and INSERM. This work received financial support from CNRS, Université Côte d'Azur, the Canceropôle PACA Research Fund, ANR, INCA, La Ligue Nationale Contre le Cancer (Equipe Labellisée 2019), Fondation ARC pour la Recherche sur le Cancer (Programme Labellisé 2022), and the ARCAENGINE2023020006332 program.

Informed Consent Statement: Since the information we used for our manuscript came from publicly available databases, informed consent was obtained from all subjects involved in the study.

Data Availability Statement: the data were from publicly available databases including the R2 Genomics Analysis and Visualization Platform https://hgserver1.amc.nl/cgi-bin/r2/main.cgi?open_page=login and the TCGA databases from the cbiportal visualization platform <https://www.cbiportal.org/>.

Acknowledgments: We thank the Fondation Flavien for its continuous support to our program <https://www.fondationflavien.com/>.

Conflicts of Interest: The authors declare no conflicts of interest.

Abbreviations

The following abbreviations are used in this manuscript:

ABL	Abelson Tyrosine-Protein Kinase 1
ALK	Anaplastic Lymphoma Kinase
AXL	Tyrosine-protein kinase receptor UFO
BCL2	B Cell Lymphoma 2
BCR	Breakpoint Cluster Region protein
BRAF	B-Raf protein, serine/threonine kinase
BTK	Bruton Kinase
CDK	Cyclin Dependent Kinase
CTLA4	Cytotoxic T-Lymphocyte Associated Protein 4
DDR2	Discoidin Domain Receptor Tyrosine Kinase 2
DFS	Disease Free Survival
EGFR/HER1	Epithelial Growth Factor Receptor
EPCAM	Epithelial Cell Adhesion Molecule
FDA	Food and Drug Administration
FGF	Fibroblast growth factors
FILP1	Filamin A Interacting Protein 1
FLT3	Fms Related Receptor Tyrosine Kinase 3
HGFR/MET	Hepatocyte Growth Factor Receptor
HRP	Horse Radish Peroxidase
KIT	KIT proto-oncogene, receptor tyrosine kinase
IDH1	Isocitrate dehydrogenase
JAK	Janus Kinase
LAG 3	Lymphocyte Activating 3
LYN	Protein tyrosine kinase Src family
MAPK	Mitogen-activated protein kinase
MEK	Mitogen Activated Protein Kinase Kinase
mTOR	Mammalian target of rapamycin
OS	Overall Survival
PARP	Poly ADP Ribose Polymerase
PD1/PDL1	Programmed Cell Death Protein 1/PD1 Ligand
PDGFR	Platelet Derived Growth Factor Receptor
PFS	Progression Free Survival
PKC	Protein Kinase C
PI	Propidium Iodide

PI3K	Phosphoinositide 3-kinase
PVDF	Polyvinylidene Fluoride
RAF	Rapidly Accelerated Fibrosarcoma
RET	Rearranged during transfection tyrosine kinase receptor
ROS	ROS Proto-Oncogene 1, Receptor Tyrosine Kinase
SMO	Smoothed protein
TIM3	T-cell immunoglobulin and mucin-domain containing-3
TRK	Tropomyosin Receptor Kinase
VEGFR	Vascular Endothelial Growth Factor Receptor

References

- Gajjar, A.; Bowers, D.C.; Karajannis, M.A.; Leary, S.; Witt, H.; Gottardo, N.G. Pediatric Brain Tumors: Innovative Genomic Information Is Transforming the Diagnostic and Clinical Landscape. *J Clin Oncol* **2015**, *33*, 2986-2998, doi:10.1200/JCO.2014.59.9217.
- Gajjar, A.; Chintagumpala, M.; Ashley, D.; Kellie, S.; Kun, L.E.; Merchant, T.E.; Woo, S.; Wheeler, G.; Ahern, V.; Krasin, M.J.; et al. Risk-adapted craniospinal radiotherapy followed by high-dose chemotherapy and stem-cell rescue in children with newly diagnosed medulloblastoma (St Jude Medulloblastoma-96): long-term results from a prospective, multicentre trial. *Lancet Oncol* **2006**, *7*, 813-820, doi:10.1016/S1470-2045(06)70867-1.
- Packer, R.J.; Gajjar, A.; Vezina, G.; Rorke-Adams, L.; Burger, P.C.; Robertson, P.L.; Bayer, L.; LaFond, D.; Donahue, B.R.; Marymont, M.H.; et al. Phase III study of craniospinal radiation therapy followed by adjuvant chemotherapy for newly diagnosed average-risk medulloblastoma. *J Clin Oncol* **2006**, *24*, 4202-4208, doi:10.1200/JCO.2006.06.4980.
- Coy, S.; Lee, J.S.; Chan, S.J.; Woo, T.; Jones, J.; Alexandrescu, S.; Wen, P.Y.; Sorger, P.K.; Ligon, K.L.; Santagata, S. Systematic characterization of antibody-drug conjugate targets in central nervous system tumors. *Neuro Oncol* **2024**, *26*, 458-472, doi:10.1093/neuonc/noad205.
- Thompson, E.M.; Keir, S.T.; Venkatraman, T.; Lascola, C.; Yeom, K.W.; Nixon, A.B.; Liu, Y.; Picard, D.; Remke, M.; Bigner, D.D.; et al. The role of angiogenesis in Group 3 medulloblastoma pathogenesis and survival. *Neuro Oncol* **2017**, *19*, 1217-1227, doi:10.1093/neuonc/nox033.
- Rini, B.I.; Escudier, B.; Tomczak, P.; Kaprin, A.; Szczylik, C.; Hutson, T.E.; Michaelson, M.D.; Gorbunova, V.A.; Gore, M.E.; Rusakov, I.G.; et al. Comparative effectiveness of axitinib versus sorafenib in advanced renal cell carcinoma (AXIS): a randomised phase 3 trial. *Lancet* **2011**, *378*, 1931-1939, doi:10.1016/S0140-6736(11)61613-9 [pii] 10.1016/S0140-6736(11)61613-9.
- Rini, B.I.; Plimack, E.R.; Stus, V.; Gafanov, R.; Hawkins, R.; Nosov, D.; Pouliot, F.; Alekseev, B.; Soulieres, D.; Melichar, B.; et al. Pembrolizumab plus Axitinib versus Sunitinib for Advanced Renal-Cell Carcinoma. *N Engl J Med* **2019**, *380*, 1116-1127, doi:10.1056/NEJMoa1816714.
- Pagnuzzi-Boncompagni, M.; Picco, V.; Vial, V.; Planas-Bielsa, V.; Vandenberghe, A.; Daubon, T.; Derieppe, M.A.; Montemagno, C.; Durivault, J.; Grepin, R.; et al. Antiangiogenic Compound Axitinib Demonstrates Low Toxicity and Antitumoral Effects against Medulloblastoma. *Cancers (Basel)* **2021**, *14*, doi:10.3390/cancers14010070.
- Schwinn, S.; Mokhtari, Z.; Thusek, S.; Schneider, T.; Siren, A.L.; Tiemeyer, N.; Caruana, I.; Miele, E.; Schlegel, P.G.; Beilhack, A.; et al. Cytotoxic effects and tolerability of gemcitabine and axitinib in a xenograft model for c-myc amplified medulloblastoma. *Sci Rep* **2021**, *11*, 14062, doi:10.1038/s41598-021-93586-x.
- Donze, C.; Revon-Riviere, G.; Pondrom, M.; Verschuur, A.; Leblond, P.; Andre, N. Retrospective experience of children with relapsed brain tumors treated with oral combination of axitinib and metronomic etoposide. *Pediatr Blood Cancer* **2024**, *71*, e31076, doi:10.1002/pbc.31076.
- Ahmed, N.; Ratnayake, M.; Savoldo, B.; Perlaky, L.; Dotti, G.; Wels, W.S.; Bhattacharjee, M.B.; Gilbertson, R.J.; Shine, H.D.; Weiss, H.L.; et al. Regression of experimental medulloblastoma following transfer of HER2-specific T cells. *Cancer Res* **2007**, *67*, 5957-5964, doi:10.1158/0008-5472.CAN-06-4309.

12. Cavalli, F.M.G.; Remke, M.; Rampasek, L.; Peacock, J.; Shih, D.J.H.; Luu, B.; Garzia, L.; Torchia, J.; Nor, C.; Morrissy, A.S.; et al. Intertumoral Heterogeneity within Medulloblastoma Subgroups. *Cancer Cell* **2017**, *31*, 737-754 e736, doi:10.1016/j.ccell.2017.05.005.
13. Ho, D.M.; Shih, C.C.; Liang, M.L.; Tsai, C.Y.; Hsieh, T.H.; Tsai, C.H.; Lin, S.C.; Chang, T.Y.; Chao, M.E.; Wang, H.W.; et al. Integrated genomics has identified a new AT/RT-like yet INI1-positive brain tumor subtype among primary pediatric embryonal tumors. *BMC Med Genomics* **2015**, *8*, 32, doi:10.1186/s12920-015-0103-3.
14. Robinson, G.; Parker, M.; Kranenburg, T.A.; Lu, C.; Chen, X.; Ding, L.; Phoenix, T.N.; Hedlund, E.; Wei, L.; Zhu, X.; et al. Novel mutations target distinct subgroups of medulloblastoma. *Nature* **2012**, *488*, 43-48, doi:10.1038/nature11213.
15. de Bont, J.M.; Kros, J.M.; Passier, M.M.; Reddingius, R.E.; Sillevius Smitt, P.A.; Luijck, T.M.; den Boer, M.L.; Pieters, R. Differential expression and prognostic significance of SOX genes in pediatric medulloblastoma and ependymoma identified by microarray analysis. *Neuro Oncol* **2008**, *10*, 648-660, doi:10.1215/15228517-2008-032.
16. Kool, M.; Koster, J.; Bunt, J.; Hasselt, N.E.; Lakeman, A.; van Sluis, P.; Troost, D.; Meeteren, N.S.; Caron, H.N.; Cloos, J.; et al. Integrated genomics identifies five medulloblastoma subtypes with distinct genetic profiles, pathway signatures and clinicopathological features. *PLoS One* **2008**, *3*, e3088, doi:10.1371/journal.pone.0003088.
17. Kool, M.; Jones, D.T.; Jager, N.; Northcott, P.A.; Pugh, T.J.; Hovestadt, V.; Piro, R.M.; Esparza, L.A.; Markant, S.L.; Remke, M.; et al. Genome sequencing of SHH medulloblastoma predicts genotype-related response to smoothed inhibition. *Cancer Cell* **2014**, *25*, 393-405, doi:S1535-6108(14)00073-7 [pii] 10.1016/j.ccr.2014.02.004.
18. Fattet, S.; Haberler, C.; Legoix, P.; Varlet, P.; Lellouch-Tubiana, A.; Lair, S.; Manie, E.; Raquin, M.A.; Bours, D.; Carpentier, S.; et al. Beta-catenin status in paediatric medulloblastomas: correlation of immunohistochemical expression with mutational status, genetic profiles, and clinical characteristics. *J Pathol* **2009**, *218*, 86-94, doi:10.1002/path.2514.
19. Roth, R.B.; Hevezi, P.; Lee, J.; Willhite, D.; Lechner, S.M.; Foster, A.C.; Zlotnik, A. Gene expression analyses reveal molecular relationships among 20 regions of the human CNS. *Neurogenetics* **2006**, *7*, 67-80, doi:10.1007/s10048-006-0032-6.
20. Northcott, P.A.; Buchhalter, I.; Morrissy, A.S.; Hovestadt, V.; Weischenfeldt, J.; Ehrenberger, T.; Grobner, S.; Segura-Wang, M.; Zichner, T.; Rudneva, V.A.; et al. The whole-genome landscape of medulloblastoma subtypes. *Nature* **2017**, *547*, 311-317, doi:10.1038/nature22973.
21. Pugh, T.J.; Weeraratne, S.D.; Archer, T.C.; Pomeranz Krummel, D.A.; Auclair, D.; Bochicchio, J.; Carneiro, M.O.; Carter, S.L.; Cibulskis, K.; Erlich, R.L.; et al. Medulloblastoma exome sequencing uncovers subtype-specific somatic mutations. *Nature* **2012**, *488*, 106-110, doi:10.1038/nature11329.
22. Jones, D.T.; Jager, N.; Kool, M.; Zichner, T.; Hutter, B.; Sultan, M.; Cho, Y.J.; Pugh, T.J.; Hovestadt, V.; Stutz, A.M.; et al. Dissecting the genomic complexity underlying medulloblastoma. *Nature* **2012**, *488*, 100-105, doi:10.1038/nature11284.
23. Morrissy, A.S.; Garzia, L.; Shih, D.J.; Zuyderduyn, S.; Huang, X.; Skowron, P.; Remke, M.; Cavalli, F.M.; Ramaswamy, V.; Lindsay, P.E.; et al. Divergent clonal selection dominates medulloblastoma at recurrence. *Nature* **2016**, *529*, 351-357, doi:nature16478 [pii] 10.1038/nature16478.
24. Penco-Campillo, M.; Comoglio, Y.; Feliz Morel, A.J.; Hanna, R.; Durivault, J.; Leloire, M.; Mejias, B.; Pagnuzzi, M.; Morot, A.; Burel-Vandenbos, F.; et al. VEGFC negatively regulates the growth and aggressiveness of medulloblastoma cells. *Commun Biol* **2020**, *3*, 579, doi:10.1038/s42003-020-01306-4.
25. Ndiaye, P.D.; Dufies, M.; Giuliano, S.; Douguet, L.; Grepin, R.; Durivault, J.; Lenormand, P.; Glisse, N.; Mintcheva, J.; Vouret-Craviari, V.; et al. VEGFC acts as a double-edged sword in renal cell carcinoma aggressiveness. *Theranostics* **2019**, *9*, 661-675, doi:10.7150/thno.27794.
26. Carneiro, B.A.; Cavalcante, L.; Mahalingam, D.; Saeed, A.; Safran, H.; Ma, W.W.; Coveler, A.L.; Powell, S.; Bastos, B.; Davis, E.; et al. Phase I Study of Elraglusib (9-ING-41), a Glycogen Synthase Kinase-3beta Inhibitor, as Monotherapy or Combined with Chemotherapy in Patients with Advanced Malignancies. *Clin Cancer Res* **2024**, *30*, 522-531, doi:10.1158/1078-0432.CCR-23-1916.

27. Bou-Gharios, J.; Assi, S.; Bahmad, H.F.; Kharroubi, H.; Araji, T.; Chalhoub, R.M.; Ballout, F.; Harati, H.; Fares, Y.; Abou-Kheir, W. The potential use of tideglusib as an adjuvant radio-therapeutic treatment for glioblastoma multiforme cancer stem-like cells. *Pharmacol Rep* **2021**, *73*, 227-239, doi:10.1007/s43440-020-00180-5.
28. K, D.S. Capivasertib: First Approved AKT inhibitor for the Treatment of Patients with Breast Cancer. *Anticancer Agents Med Chem* **2024**, doi:10.2174/0118715206360571241126080725.
29. Varkaris, A.; Pazolli, E.; Gunaydin, H.; Wang, Q.; Pierce, L.; Boezio, A.A.; Bulku, A.; DiPietro, L.; Fridrich, C.; Frost, A.; et al. Discovery and Clinical Proof-of-Concept of RLY-2608, a First-in-Class Mutant-Selective Allosteric PI3Kalpha Inhibitor That Decouples Antitumor Activity from Hyperinsulinemia. *Cancer Discov* **2024**, *14*, 240-257, doi:10.1158/2159-8290.CD-23-0944.
30. Sarker, D.; Dawson, N.A.; Aparicio, A.M.; Dorff, T.B.; Pantuck, A.J.; Vaishampayan, U.N.; Henson, L.; Vasist, L.; Roy-Ghanta, S.; Gorczyca, M.; et al. A Phase I, Open-Label, Dose-Finding Study of GSK2636771, a PI3Kbeta Inhibitor, Administered with Enzalutamide in Patients with Metastatic Castration-Resistant Prostate Cancer. *Clin Cancer Res* **2021**, *27*, 5248-5257, doi:10.1158/1078-0432.CCR-21-1115.
31. Suh, K.J.; Ryu, M.H.; Zang, D.Y.; Bae, W.K.; Lee, H.S.; Oh, H.J.; Kang, M.; Kim, J.W.; Kim, B.J.; Mortimer, P.G.S.; et al. AZD8186 in Combination With Paclitaxel in Patients With Advanced Gastric Cancer: Results From a Phase Ib/II Study (KCSG ST18-20). *Oncologist* **2023**, *28*, e823-e834, doi:10.1093/oncolo/oyad059.
32. Matsubara, N.; Miura, Y.; Nishiyama, H.; Taoka, R.; Kojima, T.; Shimizu, N.; Hwang, J.; Ote, T.; Oyama, R.; Toyozumi, K.; et al. Phase 3 THOR Japanese subgroup analysis: erdafitinib in advanced or metastatic urothelial cancer and fibroblast growth factor receptor alterations. *Int J Clin Oncol* **2024**, *29*, 1516-1527, doi:10.1007/s10147-024-02583-3.
33. Picca, A.; Di Stefano, A.L.; Savatovsky, J.; Ducray, F.; Chinot, O.; Moyal, E.C.; Augereau, P.; Le Rhun, E.; Schmitt, Y.; Rousseaux, N.; et al. TARGET: A phase I/II open-label multicenter study to assess safety and efficacy of fexagratinib in patients with relapsed/refractory FGFR fusion-positive glioma. *Neurooncol Adv* **2024**, *6*, vdae068, doi:10.1093/noajnl/vdae068.
34. Jung, H.A.; Park, S.; Lee, S.H.; Ahn, J.S.; Ahn, M.J.; Sun, J.M. Dacomitinib in EGFR-mutant non-small-cell lung cancer with brain metastasis: a single-arm, phase II study. *ESMO Open* **2023**, *8*, 102068, doi:10.1016/j.esmoop.2023.102068.
35. Ploumaki, I.; Triantafyllou, E.; Koumprentziotis, I.A.; Karampinos, K.; Drougkas, K.; Karavolias, I.; Trontzas, I.; Kotteas, E.A. Bcl-2 pathway inhibition in solid tumors: a review of clinical trials. *Clin Transl Oncol* **2023**, *25*, 1554-1578, doi:10.1007/s12094-022-03070-9.
36. Badisa, R.B.; Darling-Reed, S.F.; Joseph, P.; Cooperwood, J.S.; Latinwo, L.M.; Goodman, C.B. Selective cytotoxic activities of two novel synthetic drugs on human breast carcinoma MCF-7 cells. *Anticancer Res* **2009**, *29*, 2993-2996.
37. Wen, X.; Lu, Y.; Li, Y.; Qi, P.; Wu, Y.; Yu, J.; Zhang, R.; Huang, Q.; Huang, P.; Hou, B.; et al. Remission rate, toxicity and pharmacokinetics of venetoclax-based induction regimens in untreated pediatric acute myeloid leukemia. *NPJ Precis Oncol* **2024**, *8*, 248, doi:10.1038/s41698-024-00740-5.
38. Schuller, U.; Schober, F.; Kretzschmar, H.A.; Herms, J. Bcl-2 expression inversely correlates with tumour cell differentiation in medulloblastoma. *Neuropathol Appl Neurobiol* **2004**, *30*, 513-521, doi:10.1111/j.1365-2990.2004.00553.x.
39. Garancher, A.; Lin, C.Y.; Morabito, M.; Richer, W.; Rocques, N.; Larcher, M.; Bihannic, L.; Smith, K.; Miquel, C.; Leboucher, S.; et al. NRL and CRX Define Photoreceptor Identity and Reveal Subgroup-Specific Dependencies in Medulloblastoma. *Cancer Cell* **2018**, *33*, 435-449 e436, doi:10.1016/j.ccell.2018.02.006.
40. Badawi, M.; Menon, R.; Place, A.E.; Palenski, T.; Sunkersett, G.; Arrendale, R.; Deng, R.; Federico, S.M.; Cooper, T.M.; Salem, A.H. Venetoclax Penetrates the Blood Brain Barrier: A Pharmacokinetic Analysis in Pediatric Leukemia Patients. *J Cancer* **2023**, *14*, 1151-1156, doi:10.7150/jca.81795.

Disclaimer/Publisher's Note: The statements, opinions and data contained in all publications are solely those of the individual author(s) and contributor(s) and not of MDPI and/or the editor(s). MDPI and/or the editor(s) disclaim responsibility for any injury to people or property resulting from any ideas, methods, instructions or products referred to in the content.