Research Article

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Pharmacoepigenetics of the role of DNA methylation in µ-opioid receptor expression in different human brain regions

Aim: Exposure to opioids has been associated with epigenetic effects. Studies in rodents suggested a role of varying degrees of DNA methylation in the differential regulation of μ -opioid receptor expression across the brain. **Methods:** In a translational investigation, using tissue acquired postmortem from 21 brain regions of former opiate addicts, representing a human cohort with chronic opioid exposure, μ -opioid receptor expression was analyzed at the level of DNA methylation, mRNA and protein. **Results & conclusion:** While high or low μ -opioid receptor expression significantly correlated with local *OPRM1* mRNA levels, there was no corresponding association with *OPRM1* methylation status. Additional experiments in human cell lines showed that changes in DNA methylation associated with changes in μ -opioid expression were an order of magnitude greater than differences in brain. Hence, different degrees of DNA methylation associated with chronic opioid exposure are unlikely to exert a major role in the region-specificity of μ -opioid receptor expression in the human brain.

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Keywords: DNA methylation • human brain • μ-opioid receptor regulation • *OPRM1* methylation

Chronic exposure to exogenous opioids has been shown to trigger epigenetic changes, including increased methylation of the µ-opioid receptor gene OPRM1 in chronic opioid-treated pain patients and opiate addicts [1] and changes in global DNA methylation following opioid treatment of human cell lines [2]. These pharmacoepigenetic effects are likely to impact on brain opioid receptor expression. However, most evidence for this has been gathered in rodents, suggesting contributions of changes in DNA methylation, alterations in the methyl-CpG-binding protein 2 (MeCP2) mediated chromatin structure and modifications of histone acetylation to different opioid receptor expression [3,4]. For example, neuronal differentiation of P19 mouse embryonal carcinoma cells and the consequent increased µ-opioid receptor expression was accompanied by a decrease in DNA methylation [4], and μ -opioid receptor expression was also induced by exposure to the demethylating agent 5'-aza-2'-deoxycytidine or to histone deacetylase inhibitors in P19 cells. In addition, in mouse brain, μ -opioid receptor expression was correlated with reduced levels of *OPRM1* DNA methylation [4]. Moreover, MeCP2 (methyl binding protein) and DNA methylation were found to mediate the remodeling of the *OPRM1* promoter via chromatin remodeling factors (Brg1 and BAF155) during neuronal differentiation [5].

Understanding of the role of opioid-triggered epigenetic mechanisms in regulating μ -opioid receptor expression in humans, in particular in a pharmacoepigenetic context, has resulted in the identification of an increasing number of epigenetic effects of common drugs [6]. In the present study, we investigated Claudia Knothe¹, Bruno G Oertel², Alfred Ultsch³, Mattias Kettner⁵, Peter Harald Schmidt⁴, Cora Wunder⁵, Stefan W Toennes⁵, Gerd Geisslinger^{1,2} & Jörn Lötsch*^{1,2}

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a possible association between differential μ -opioid receptor expression across several brain areas and varying degrees of *OPRM1* DNA methylation. Since active manipulation by opioid exposure of brain opioid receptor expression is impossible in living subjects, brain tissue was collected postmortem from heroin addicts as a cohort with long-term exposure to opioids. As a mechanistic control, the possible association of μ -opioid receptor expression with *OPRM1* DNA methylation was assessed in human-derived cell lines.

Materials & methods

Analysis of the μ -opioid receptor expression pattern across pain or addiction relevant human brain areas

Study cohort & tissue sample acquisition

The protocol was approved by the Ethics Committee of the Medical Faculty of the Goethe - University, Frankfurt am Main, Germany. Brain tissue was gathered from n = 27 Caucasian fatalities (for demographic details, see Table 1) during forensic examination. Subjects were examined for whom drug intoxication was the probable cause of death. After evaluating the circumstances of death, autopsy results, police reports, medical records (if available) and toxicological data related to opioids (blood or urine concentrations of 6-monoacetylmorphine, morphine/codeine blood concentration ratio >1 [7]), the deaths of 24 of the 27 subjects were verifiably linked to illicit drug consumption; the three additional subjects were also included in order to accommodate a wide range of possible opioid exposure. Average opioid concentrations (±SD) were 324.03 ng/g (±281.02) morphine (n = 19), 191.59 ng/g (±244.30) 6-monoacetylmorphine (n = 15), 82.14 ng/g (\pm 82.67) codeine (n = 17), 1815.67 ng/g (±1151.71) methadone (n = 6)and 18.95 ng/g (± 2.84) fentanyl (n = 1).

During the postmortem examination, a forensic pathologist removed tissue samples from n = 21 brain regions (Table 2) that can be regarded, on the basis of published evidence, as being relevant to either pain/analgesia [8-10] or opioid addiction [11,12], representing relevant clinical settings of chronic opioid exposure. These areas involved in pain and/or addiction are known to be associated with the expression of μ -opioid receptors [9,13-16]. To increase the range of μ -opioid receptor expression in the tested samples, brain areas were also sampled in which μ -opioid receptors are reportedly absent, such as the occipital cortex or the cerebellum [16,17].

Opioid receptor tissue expression assessment Membrane protein extraction from human brain tissue samples

Membrane preparations were obtained as described previously [18]. In brief, tissue (1–1.5 g) was homog-

enized with a fast rotating Ultra-Turrax (Janke & Kunkel GmbH, Staufen i. Breisgau, Germany) in 10–15 volumes of homogenization buffer (10 mM Tris-HCl, pH 7.4, 290 mM D(+)-saccharose). The crude tissue homogenate was immediately centrifuged at 1000 \times g at 4°C for 15 min, and the supernatant again at 45,000 \times g, at 4°C for 45 min. The pellet was resuspended in ice-cold isolation buffer (50 mM Tris-HCl, pH 7.4) and stored at -80°C pending further analysis. Protein concentrations were measured according to the method of Bradford [19].

Opioid agonist saturation binding

The µ-opioid receptor expression (B_{max}) was assessed by incubating membrane proteins (30 µg) with ten different concentrations, ranging from 0.06 to 13.1 nM, of the µ-opioid receptor-specific radioligand ³H-[D-Ala²,N-MePhe⁴,Gly-ol⁵]-enkephalin ([³H]-DAMGO) (Perkin Elmer, MA, USA) in the presence and absence of 10 µM naloxone. Experiments were performed at room temperature in 96-well MultiScreen_{HTS}-FB Plates (Millipore, MA, USA) containing a final volume of 200 µl assay buffer (50 mM Tris-HCl, pH 7.4). After an incubation period of 60 min, membrane bound and unbound [3H]-DAMGO was separated by rapid filtration through the glass fiber filters at the bottom of the wells using a MultiScreen_{HTS}-Vacuum Manifold (Millipore). Subsequently, the filters were washed six-times with 200 µl of ice-cold assay buffer, dried and covered with 30 µl liquid scintillation cocktail (Quicksafew A, Zinsser Analytic GmbH, Frankfurt, Germany). The decay per minute of the filter-bound [3H]-DAMGO was measured using a MicroBeta² Microplate Counter (Perkin Elmer). The number of µ-opioid receptors (B_{max}) was calculated by fitting a nonlinear model to the data: $B([^{3}H]-DAMGO) = (B_{max}([^{3}H]-DAMGO))$ $\cdot F([^{3}H]-DAMGO))/(KD([^{3}H]-DAMGO) + F([^{3}H]-$ DAMGO)) where B([³H]-DAMGO) denotes the concentration of µ-opioid receptor bound [3H]-DAMGO (specific binding), F([³H]-DAMGO) the concentration of the free [3H]-DAMGO, B_{max}([3H]-DAMGO) the maximum number of [3H]-DAMGO binding sites (µ-opioid receptor density) and KD([³H]-DAMGO) the equilibrium dissociation constant of [3H]-DAMGO. Calculations were done using the GraphPad Prism software (version 5.03 for Windows®, GraphPad Inc., CA, USA).

Analysis of the regulation of μ -opioid receptor expression in prototypic brain areas

The regulation of μ -opioid receptor expression was assessed at three areas selected on the basis of the following criteria: prototypic representation of the brain regions with the most pronounced differences

ID	Sex (m/f)	Age at death (years)	Postmortem delay (h)	Storage time (months)	Cause of death
18/09	m	24	25	71.7	Intoxication/drug overdose
38/09	f	38	15	71.7	Intoxication/drug overdose
39/09	m	30	8	71.5	Intoxication/drug overdose
86/08	m	36	79	72.6	Intoxication/drug overdose
108/09	m	49	31	71.5	Intoxication/drug overdose
135/08	f	38	20	72.6	Intoxication/pneumonia/enteritis
141/09	m	39	11	71.5	Intoxication/drug overdose/HIV
169/09	m	43	37	71.5	Intoxication/drug overdose
182/08	m	29	58	72.6	Intoxication/drug overdose
183/08	m	36	65	72.5	Intoxication/drug overdose/alcohol
235/08	m	49	15	72.4	Intoxication/drug overdose/bronchial carcinoma/HIV/ coronary heart disease
236/08	m	34	8	72.4	Intoxication/drug overdose
261/09	m	35	24		Intoxication/drug overdose
286/08	m	48	76	72.4	Intoxication/drug overdose, heart failure
316/08	m	26	14	72.0	Intoxication/drug overdose
328/08	f	42	15	72.0	Intoxication/drug overdose
364/08	m	29	25	72.2	Intoxication/drug overdose
374/08	m	42	24	72.2	Intoxication/drug overdose/hyperglycemia diabetes mellitus/alcohol
423/08	m	29	42	72.2	Intoxication/drug overdose
439/08	f	48	17	72.2	Intoxication/drug overdose
524/08	m	49	20	71.8	Intoxication triggered heart failure
704/09	m	44	33	71.3	Intoxication/drug overdose/alcohol
789/08	m	49	37	71.8	Intoxication/drug overdose
886/08	m	43	25	71.7	Intoxication/drug overdose/pathological organ report
229/08	m	49	87	72.4	Heart failure
332/08	f	26	33	72.0	Heart failure
554/10	f	86	84		Pathological organ report
Sum or mean ± SD	21/6	40.4 ± 12	34.4 ± 23.7	72.03 ± 0.4	

in expression, availability of sufficient tissue volume and acceptable quality for the necessary analysis. This resulted in the selection of two regions form the largest cluster that also displayed the highest μ -opioid receptor expression, namely, the thalamus medialis and the basolateral amygdala, as well as one region that displayed the lowest *OPRM1* expression, namely, the nucleus pontis. In these areas, regulation of μ -opioid receptor expression was assessed. In these experiments, the hypothesis that DNA methylation plays a role in the regulation of μ -opioid receptor expression was investigated by mRNA quantification and DNA methylation analysis. In addition, experiments that could not be carried out on the tissue samples were performed in human cell lines as described below.

Quantification of OPRM1 mRNA expression

The μ -opioid receptor gene mRNA expression was analyzed using a validated TaqMan gene expression assay (Applied Biosystems, Darmstadt, Germany), as described previously with slight modifications [20]. In brief, RNA was isolated from tissue samples stored at -80°C of thalamus medialis, basolateral amygdala and nucleus pontis or cell pellets stored at -80°C by means of the RNeasy Mini Kit (Qiagen). To avoid the degradation of contained mRNA during thawing, samples were defrosted in appropriate lysis buffer included in the kit. Isolation of mRNA was done as recommended by the manufacturer including DNase digestion to eliminate genomic DNA contamination. Finally, mRNA was eluted with RNase-free water and stored at -80°C until further use. RNA extraction was done promptly prior cDNA synthesis to avoid degradation during storage at -80°C. The quality and quantity of the extracted RNA was verified using the NanoQuant technology of the Tecan infinite F200 PRO (Tecan Trading AG, Switzerland). Subsequently, a reverse transcription into DNA was performed using 500 ng of extracted RNA and random primers included in the First Strand cDNA Synthesis Kit (Thermo ScientificTM, Schwerte, Germany) according to the manufacturer's protocol. Real-time polymerase chain reaction (PCR) analyses were done with 1 µl cDNA template in triplicates using the 2× TaqMan Gene Expression Mastermix and the validated TagMan gene expression assay for the OPRM1 gene (Hs01053957_ m1, Applied Biosystems) and the TaqMan endogenous control assays for human β-actin (Hs99999903_m1) and for human glyceraldehyde-3-phosphate dehydrogenase (GAPDH; Hs03929097_g1). For cell samples, previously published primers for the endogenous controls were used: human B-actin: for 5'-TGACCCA-GATCATGTTTGAG-3', rev 5'-TTAATGT-CACGCACGATTTCC-3' and human GAPDH: for 5'-GATTCCACCCATGGCAAATTC-3' rev 5'-TGGGATTTCCATTGATGACAAG3' [21]. Realtime-PCR for baseline determination of endogenous OPRM1 expression in SHSY5Y, Kelly and HEK-293 cells were done with 1 µl cDNA template using the 2× ABsolute Blue QPCR Mix, SYBR Green, ROX (Thermo ScientificTM, Schwerte, Germany) for β -actin and GAPDH and with 2× TaqMan Gene Expression Mastermix for the OPRM1 gene. To increase quantification of OPRM1 expression in Kelly and HEK-293 cells following 5-Aza-2-desoxy-cytidine (5-Aza-CdR) treatment 2 µl of cDNA template were used. Quantification of mRNA was conducted on a 7500 Fast-Real-Time PCR System (Applied Biosystems). Using the implemented 7500 software v.2.0.1. Relative OPRM1 mRNA expression was determined by the comparative Ct method. OPRM1 mRNA transcripts were normalized to GAPDH and finally related to β-actin when

Table 2. Brain regions, ³H-[D-Ala²,N-MePhe⁴,Gly-ol⁵]-enkephalin B_{Max} as a measure of μ -opioid binding sites and cluster membership of the investigated brain membrane preparations, sorted for decreasing median μ -opioid receptor expression level (compare Figure 1B).

Brain region	Median B _{max} (fmol/mg protein)	(Interquartile range)	n	Cluster					
Thalamus medialis	56.30	(29.9–80.3)	26	1					
Basolateral amygdala	53.80	(36.3–68.9)	26	1					
SI	39.40	(19.7–55.4)	24	1					
Thalamus lateralis	35.60	(24–50.7)	24	1					
Orbitofrontal cortex	34.10	(14.5–41.2)	27	1					
Nucleus accumbens	32.90	(18.4–56)	26	1					
Insula	30.50	(22.9–48.7)	26	1					
Prefrontal cortex	29.30	(22.9–39.2)	27	1					
Anterior cingulum	28.40	(19.3–46.5)	27	1					
Striatum: nucleus caudatus	27.50	(15–34.6)	25	1					
SII	27.40	(22.7–39.4)	22	1					
Spinal cord	27.20	(12.2–38.8)	22	4					
Hypothalamus	26.80	(16.9–32.3)	26	1					
Striatum: putamen	26.30	(17–43.4)	26	1					
Cerebellum vermis	19.70	(5.8–27.4)	22	4					
Tegmentum pontis	18.10	(4.8–34)	22	2					
Cerebellum hemisphere	15.90	(4.1–25)	24	2					
Medulla oblongata	11.20	(5.1–22.3)	20	2					
Hippocampus	8.90	(1.8–18.9)	23	2					
Occipital cortex	2.30	(0.4–11.7)	22	3					
Nucleus pontis	2.00	(0–12.5)	21	3					

expression levels were compared between different cell types or brain regions. In case of comparisons between untreated cells and 5-Aza-CdR treated cells *OPRM1* mRNA transcripts were normalized to GAPDH and finally related to the control condition of the SHSY5Y cell line.

Quantification of OPRM1 DNA methylation

To analyze possible regulation of the µ-opioid receptor brain expression byDNA methylation, the degree of DNA methylation at CpG islands in the OPRM1 gene region was measured using a validated PyrosequencingTM assay as described previously [1]. This agrees with contemporary laboratory assays in the present context [20,22-23]. A total of 22 CpG positions (from position 93 to +159) within a CpG-island reaching from position -382 to +561 relative to the start codon was analyzed. In brief, genomic DNA was extracted from tissue preparations or cell pellets, which were stored at -80°C using the DNeasy blood and tissue Kit (Qiagen, Hilden, Germany) according to manufacturer's protocol. Bisulphite treatment was then performed with 500 ng DNA using the EZ DNA Methylation-Gold Kit (Zymo Research, CA, USA) according to manufacturer's instructions and eluted in 30 µl elution buffer. PCRs were run on a Mastercycler nexus gradient flexlid device (Eppendorf, Hamburg, Germany) a 25 µl reaction volume including 2.5 µl bisulphite-treated DNA, mixed with 0.25 µl MyTaqTM HS DNA Polymerase (5 U/µl) (Bioline, Luckenwalde, Germany), 5 µl 5× MyTaq Reaction Buffer, 0.1 µl of each PCR primer (100 µM) and 17.05 µl HPLC-purified water. Nested PCR provided a final PCR product of 316 bp (for primer sequences see Supplementary methods). However, due to the known susceptibility to contamination, the assay was subsequently modified toward a non-nested PCR. This modification resulted in two PCR products covering the CpG positions from -93 to +53 (sequence 1, 227 bp) and from +84 to +159 (sequence 2, 153 bp). PCR was performed with an initial denaturation step for 3 min at 92°C, and 40 cycles with a 15 s denaturation step at 92°C, an 15 s annealing step (temperature see Supplementary methods) and an 15 s elongation step at 72°C, with final elongation at 72°C for 5 min. Analysis of the OPRM1 methylation was done by means of PyrosquencingTM (Qiagen, Hilden, Germany) as described previously [1,24-25]. In brief, 25 µl of the PCR templates were processed and purified with the PyroMark Vacuum Prep Worktable (Biotage, Uppsala Sweden) and subsequently annealed to the sequencing primer (see Supplementary methods) at 80°C for 2 min as instructed by the manufacturer.

Sequence analysis took place on a PSQ 96 MA System using the PyroMark Gold Q96 Reagents (Qiagen, Hilden, Germany) with the Pyro Q-CpG methylation software (version 1.0.9) which was used to determine the nucleotide dispensation order. The software also automatically analyzed the methylation results for each single CpG site in the analyzed sequence. Each brain sample was analyzed in triplicate within one batch and two independent measurements with separate PCR amplification were conducted. For assay validation, each run included control DNA samples obtained from the EpiTect PCR Control DNA Set (Qiagen) that contained both bisulphite converted completely methylated as well as unmethylated DNA, as positive controls, and unconverted unmethylated DNA, as negative control. In the measurements using nested PCR, the average methylation over all 22 CpG sites of the unmethylated control was $3.07 \pm 0.39\%$ and the methvlated control reached 88.34 ± 0.59%. Corresponding results were validated for four CpG sites (+12 to +53) by using alternative primer pairs without nested PCR to exclude amplification bias (40 cycles instead of 65 cycles with nested PCR). To this end, samples from different brain regions were randomly composed on each 96-well plate to minimize batch effects. The same EpiTect PCR Control DAN Set was used as internal control and verified the unmethylated control as 4.60 ± 0.27% and the methylated control as 85.45 ± 1.43 %, indicating that the cycle amount of PCR amplification did not affect obtained results. In the human brain samples, comparable results were obtained for CpG sites +12 to +53 using either nested PCR or the alternative primers without nested PCR (unmethylated DNA: 7.06 ± 3.25%, methylated DNA: 87.26 ± 5.76).

Cell line based assessment of the regulatory importance of DNA methylation for *OPRM1* expression

The possibility of epigenetic regulation in humans was assessed using available human cell lines that differed in µ-opioid receptor expression. Specifically, SHSY5Y and Kelly cells are both neuroblastoma cell lines that differ in the extent of their µ-opioid expression. SHSY5Y cells express µ-opioid receptors endogenously [26], whereas Kelly cells do not express µ-opioid receptors. The non-neuronal cell line, HEK-293 (human embryonal kidney cells), that does not express µ-opioid receptors was also studied [26]. SHSY5Y and Kelly cells were selected because of their neuronal character, thereby representing a model cell system for neuronal cells. HEK-293 cells, in contrast, were chosen because of their different tissue of origin and the existing evidence for the differential regulatory impact of DNA methylation on µ-opioid receptor expression in neuronal and non-neuronal tissue [26]. Alterations in OPRM1 mRNA expression were investigated following exposure of human cell lines to the demethylating agent, 5-Aza-CdR. Since HEK-293 cells show a decreased response to the cell cycle dependent 5-Aza-CdR treatment, due to increased division rate and their reduced vulnerability compared with SHSY5Y cells and Kelly cells, 5-Aza-CdR treatment in HEK-293 cells was carried out at a higher concentration (1 μ M) to achieve a comparable decrease in DNA methylation. SHSY5Y and Kelly cells, in turn, are more vulnerable and die when treated with the cytotoxic concentration of 1 μ M 5-Aza-CdR.

HEK-293 cells were cultured in Dulbecco's modified Eagle's Medium (DMEM) + GlutaMaxTM (Gibco, Darmstadt, Germany) supplemented with 10% (v/v) fetal calf serum (FCS) and 1% penicillin/streptomycin (PAA Laboratories GmbH, Cölbe, Germany). Kelly cells were cultured in Roswell Park Memorial Institute medium (RPMI) (Gibco), supplemented with 10% (v/v) FCS and 1% penicillin/streptomycin (PAA Laboratories GmbH). SHSY5Y cells were obtained from the DZMS-Collection of Microorganisms and Cell Cultures (DSMZ, Braunschweig, Germany) and were grown in 1:1 mixed Ham F12 and Minimum Essential Medium (Gibco) supplemented with 15% (v/v) FCS, 1 % Minimum Essential Medium Non-Essential Amino Acids (Gibco), 2 mM l-glutamine (Gibco) and 1% penicillin/streptomycin (PAA Laboratories GmbH). All cells were grown at 37°C in humidified atmosphere containing 5% CO2. Prior to drug treatment, cells were seeded at an appropriate density in 10 cm² cell culture dishes and allowed to settle for 24 h in complete media. Subsequently, cells were incubated for 72 h in the presence of 0.3 or 1 μ M 5-Aza-CdR. DMSO served as vehicle control.

Data analysis

Digital pattern analysis of $\mu\text{-}opioid$ receptor expression among human brain areas

Unsupervised computer learning using a probabilistic approach [27] was employed to identify patterns of μ -opioid brain expression. This aimed at identifying structures within the data space D described by features X R²¹, which presently consisted of the quantitative receptor expressions observed in 21 different brain regions. Following visual inspection of the distribution of the B_{max} values displayed as probability density function, estimated by means of the pareto density estimation [28], data were log-transformed (zero invariant) to $LogB_{max} = ln(B_{max}+1)$ [29]. An outlier was eliminated based on a significant Grubbs test (p < 2.2×10^{-16}). Missing data (n = 79 items from the 21×27 brain regions vs tissue matrix equaling 13% of this matrix) were calculated using a k nearest neighbor algorithm with k = 3 [30]. To identify a μ -opioid receptor expression pattern across the brain-regions, the 21×27 sized input space D, comprising the LogB_{max} values, was projected onto a 2D toroid grid [31] of so-called neurons with 50 rows and 82 columns (n = 4100 units). Each neuron holds, in addition to the input vector from the 21D space (brain regions), a further vector carrying 'weights' of the same dimensions as the input dimensions. The weights were initially drawn randomly from the range of the data variables. Subsequently, they were adapted to the data (learning phase, 20 epochs).

After learning was complete, data from the trained emergent self-organizing map (ESOM) were presented on the 2D toroid map. On this map, the localizations of the so-called 'best-matching units' were used to detect cluster structures on top of the neuron grid using a so-called U-Matrix, which displays the distances in the high-dimensional space as heights [32,33]. This was enhanced by calculating in addition a P-matrix [31] that displays the point density $p(x) = |\{data \text{ points } x |\}$ d(x,x) < = r estimated as the number of data points in a sphere with radius r around x at each grid point on the ESOM's output grid. The U*-matrix combines distance structures (U-matrix) and density structures (P-matrix) into a single matrix (U*-matrix) [31]. On the U*-matrix, a geographical map analogy using watersheds was used to enhance visual detection of borders of data clusters. This approach represents a topologypreserving mapping of high-dimensional data points onto a 2D grid of neurons which was therefore favored in comparison to more common clustering algorithms such as k-means, Ward, complete and average linkage [33]. Computer learning was performed using our R library Umatrix (M Thrun, F Lerch, Marburg, Germany [34,35]) with the freely available R software package (version 3.2.3 for Linux [36,37]) running on Ubuntu Linux 14.04.4 64-bit. The clusters obtained were compared with respect to µ-opioid receptor expression using Kruskal-Wallis tests [38] followed by Dunn's post hoc pairwise multiple comparisons tests [39].

Inference statistics

In brain tissue derived data, group comparisons were performed using Kruskal–Wallis tests [38] followed by Dunn's *post hoc* pairwise multiple comparisons tests [39], or using Mann–Whitney U tests in the case of only two groups. In cell culture derived data, *OPRM1* DNA methylation was compared among cell lines and treatments by means of analysis of variance for repeated measures, with 'CpG position' as within-subject factor and 'treatment' and 'cell line' as between-subject factors. *Post hoc* comparisons were done using Student's t-tests [40] or one-way ANOVA in case of more groups. Further statistical calculations included rank correlation analyses (Spearman's ρ [41]). The α -level was set at 0.05 and corrected for multiple testing using according to Bonferroni [42]. These analyses were performed using the SPSS software package (version 23 for Windows, IBM SPSS Statistics, IL, USA).

Results

Expression pattern of µ-opioid receptors

Opioid receptor expression differed across brain regions (Figure 1). ESOM projection of the regional μ -opioid receptor expression, in other words, of the log-transformed B_{max}-values, assessed by determining the saturation binding of [³H]-DAMGO to membrane preparations (Figure 1), and subsequent visualization of the cluster distances, provided a U*-matrix (Figure 2A) on which regions that were located close together and which shared similar degrees of μ -opioid receptor expression were identifiable. The result was the identification of four clusters that differed significantly (insert of Figure 2) with respect to μ -opioid receptor expression (Kruskal–Wallis $\chi^2 = 113.21$, df = 3, $p < 2.2 \times 10^{-16}$).

The thalamus medialis and the basolateral amygdala were contained in the cluster with the highest µ-opioid receptor expression (Table 2), and which comprised the largest number of brain regions. At the opposite extreme, a small cluster was also identified comprising only the nucleus pontis and the occipital cortex. Two regions with high µ-opioid receptor expression (thalamus medialis, basolateral amygdala) were selected from the largest cluster and one region with low µ-opioid receptor expression (nucleus pontis) was selected from the smallest cluster for further analyses. These regions met the predefined criteria of prototypic representation of extremely differing clusters together with the availability of sufficient tissue volume of acceptable quality. The expression of µ-opioid receptors differed statistically significantly (Figure 1A) as expected from the cluster membership (Kruskal–Wallis $\chi^2 = 34.99$, df = 2, $p < 2.5 \times 10^{-8}$), in other words, the thalamus medialis and the basolateral amygdala did not differ from each other but both differed significantly from the nucleus pontis (Dunn's tests: p < 0.0001).

Epigenetic regulation of μ -opioid receptor expression

Assessments in human brain tissue

In the three brain areas selected for the assessments of an association between *OPRM1* DNA methylation and μ -opioid receptor expression, statistically significant differences in the amount of *OPRM1* mRNA were observed (Kruskal–Wallis $\chi^2 = 34.81$, df = 2, p < 2.8 × 10⁻⁸). The direction of the differences between brain regions agreed with the expectation, in other words, regions with high μ -opioid receptor protein expression (thalamus medialis, basolateral amygdala) contained higher *OPRM1* mRNA levels than the region with low μ -opioid receptor protein expression (nucleus pontis; Figure 3B; Dunn's tests: p < 0.001).

However, subsequent analyses of OPRM1 DNA methylation failed to provide support for a biologically plausible epigenetic regulation of µ-opioid receptor expression. The average OPRM1 DNA methylation ranged from 2.87 ± 2.01 in the thalamus medialis to $4.22 \pm 3.11\%$ in the basolateral amygdala (Figure 3C). Although the statistical analysis of the interregional differences produced a significant result (Kruskal-Wallis χ^2 = 11.13, df = 2, p = 0.0037), the most pronounced difference was observed within the cluster of high receptor expression (Dunn's tests: p = 0.004 for thalamus medialis vs nucleus pontis, other pairwise comparisons not significant). As OPRM1 mRNA expression or µ-opioid receptor expression were also not significantly correlated with OPRM1 methylation (Spearman's correlation analysis: p > 0.05), the hypothesis of epigenetic regulation had to be rejected.

Analysis in human cell lines

The possibility of epigenetic regulation of human μ -opioid receptor expression was additionally analyzed in human cell lines selected for their differences in μ -opioid receptor expression, comprising the μ -opioid receptor expressing neuronal-derived SHSY5Y cell line, the neuronal μ -opioid receptor nonexpressing Kelly cell line and the non-neuronal μ -opioid receptor nonexpressing HEK-293 cell line.

The amounts of *OPRM1* mRNA differed significantly among the three cell lines (Figure 4B) in agreement with known patterns [26] of endogenous μ -opioid receptor expression (one-way ANOVA df =2, F = 63.72, p = 0.91 × 10⁻⁴; *post hoc t*-test with Bonferroni α correction: p < 0.001 for SHSY5Y vs Kelly and SHSY5Y vs HEK-293).

Consistent with the cell-line specific mRNA expression levels, the methylation status of the *OPRM1* gene differed among the cell lines. Specifically, the analyzed CpG island including 22 CpG positions was methylated in the μ -opioid receptor nonexpressing neuronal Kelly and non-neuronal HEK-293 cells, whereas *OPRM1* methylation was almost absent in the μ -opioid receptor expressing neuronal SHSY5Y cells (Figure 4A & C). These differences were statistically significant (rm-ANOVA main effect 'cell line,' df = 2, F = 192.2, p = 2.04 × 10⁻¹¹ and 'CpG position,' df = 20, F = 28.015, p = 1.4 × 10⁻⁵, interaction 'cell line' by 'CpG position,' df = 40, F = 13.455, p = 9.5 × 10⁻⁵).

To establish a causal relationship between *OPRM1* methylation and mRNA expression, cell lines were

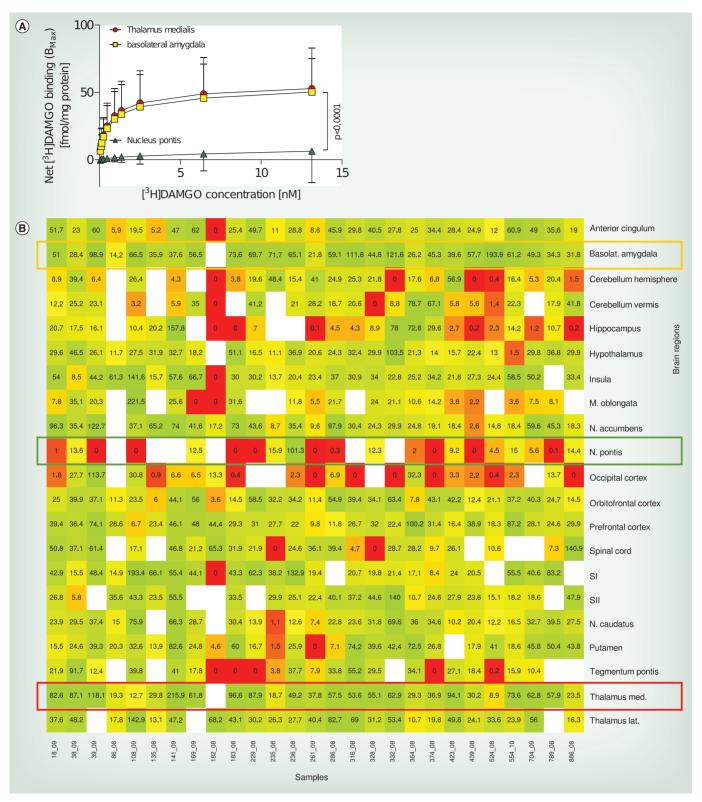


Figure 1. Expression of μ -opioid receptors across 21 different human brain regions, analyzed from tissue acquired postmortem from 27 former heroin addicts. (A) [³H]-DAMGO saturation binding at the three selected brain areas (framed in the matrix plot below) thalamus medialis, basolateral amygdala and nucleus pontis (means and SDs). The number of μ -opioid receptors (B_{max}) was calculated by fitting a nonlinear model (lines) to the data as: $B([^{3}H]DAMGO) = (B_{max}([^{3}H]DAMGO) \cdot F([^{3}H]DAMGO))/(KD([^{3}H]DAMGO) + F([^{3}H]DAMGO)) + F([^{3}H]DAMGO))$, where $B([^{3}H]DAMGO)$ denotes the concentration of μ -opioid receptor bound [³H]DAMGO (specific binding),

Figure 1. Expression of μ -opioid receptors across 21 different human brain regions, analyzed from tissue acquired postmortem from 27 former heroin addicts (cont.). (A, cont.) F([³H]DAMGO) the concentration of the free [³H]DAMGO, B_{max}([³H]DAMGO) the maximum number of [³H]DAMGO binding sites (μ -opioid receptor density) and KD([³H]DAMGO) the equilibrium dissociation constant of [³H]DAMGO. (B) Matrix plot of the B_{max} values indicating μ -opioid receptor expression observed in the 21 brain areas of 27 subjects. Cells are color coded from red to yellow according to opioid receptor expression, with red indicating lowest and yellow indicating highest expression and white cells indicate missing data.

([³H]DAMGO): ³H-[D-Ala²,N-MePhe4,Gly-ol⁵]-enkephalin; B([³H]-DAMGO): The concentration of μ -opioid receptor bound [3H]-DAMGO (specific binding); B_{max}([³H]-DAMGO): The maximum number of [³H]-DAMGO binding sites (μ -opioid receptor density); F([³H]-DAMGO): The concentration of the free [³H]-DAMGO; KD([³H]-DAMGO): The equilibrium dissociation constant of [³H]-DAMGO; B_{max}: The number of μ -opioid receptors.

treated for 3 days with the demethylating agent 5-Aza-CdR. This induced the desired decrease in the average OPRM1 methylation (Figure 5C), in other words, in Kelly cells by 30.1% following treatment with 0.3 µM 5-Aza-CdR (p = 4.01×10^{-9} vs control) and in HEK-293 cells by 10.4 % or 12.9 % following treatment with 0.3μ M or 1μ M 5-Aza-CdR, respectively (p = 0.001 and $p = 3.1 \times 10^{-5}$ vs control). In contrast, 5-Aza-CdR treatment slightly increased the already weak OPRM1 methylation in the SHSY5Y cells (p = 0.0245 vs control), which was not significant when assessed by an rm-ANOVA which included each single CpG position. In addition, the weak methylation had no impact on OPRM1 mRNA expression (Figure 5B & C). Following these treatments, OPRM1 mRNA expression in Kelly (unpaired t-test p < 0.001) and HEK-293 cells (oneway ANOVA df = 2, F = 16.8, p = 0.0003; post hoc t-test with Bonferroni α correction: p < 0.001 for 1 μ M 5-Aza-CdR) was found to be significantly upregulated (Figure 5B).

Discussion

The present observations in human brain tissue, acquired postmortem from former heroin addicts, do not support a major role of DNA methylation in the differential expression of the µ-opioid receptor across human brain regions. Over a wide range of µ-opioid receptor expression across human brain regions, the associated methylation of the OPRM1 gene displayed only a small variability. Brain areas at the extreme ends of the receptor expression range differed only by 1-2% with respect to the OPRM1 methylation, which in addition was not significantly correlated with the receptor expression. In contrast, human cell lines with different µ-opioid receptor expression differed by 30-80% with respect to the OPRM1 methylation, suggesting the possibility that OPRM1 expression in humans can be regulated via DNA methylation. However, changes in DNA methylation exceeded those observed in human brain tissue by an order of magnitude. The present data, though, do not rule out a major role of DNA methylation on µ-opioid expression in different human tissue at all.

The activation of μ -opioid receptor expression following demethylation treatment [3,4] could nevertheless, be translated to human cells in an *in vitro* setting. The µ-opioid receptor positive human neuronal cell lines are hypomethylated at a CpG island located close to the OPRM1 start codon. In contrast, µ-opioid receptor negative human cell lines are hypermethylated at this gene location. These observations are in line with recent published evidence [26,44] supporting a regulatory activity of the selected CpG island at positions -93 to +159 of the OPRM1 gene. Furthermore, the initiation of µ-opioid receptor expression following DNA demethylation with 5-Aza-CdR in Kelly and HEK-293 cells supports a possible role of DNA methylation in the expression of human µ-opioid receptors. However, even pronounced OPRM1 demethylation in formerly µ-opioid receptor negative human cell lines was unable to raise the expression level to that observed in µ-opioid receptor positive cell lines. This suggests the involvement of other mechanisms of regulation of µ-opioid receptor expression. This possibility was also suggested by an observation in µ-opioid receptor negative Jurkat cell line where receptor upregulation following demethylation (5-Aza-CdR) combined with histone deacetylase inhibition (TSA) could not reach expression levels seen in μ -opioid receptor positive cells [44].

In contrast to the experiments in cell lines, the regulatory role of DNA methylation in µ-opioid receptor expression could not be translated from mice [3] to humans. The present results on human brain samples are in contrast to data from laboratory animals that suggested a role of µ-opioid receptor gene methylation in regulation of gene expression [3,4]. In µ-opioid receptor negative P19 mouse embryonal carcinoma cells, the expression of µ-opioid receptors could be increased by neuronal differentiation, while the proximal promoter of the OPRM1 gene reverted from a highly methylated state in undifferentiated cells to hypomethylation in differentiated cells. In contrast, the methylation of other regions was not changed, and µ-opioid receptor expression could also be induced by a demethylating agent (5-Aza-CdR) and by histone deacetylase inhibitors (trichostatin A, valproic acid), suggesting the involvement of DNA methylation or histone deacetylation in µ-opioid receptor silencing. Differentiation of P19 cells reduced an interaction of the MeCP2 with the promoter region, which is relevant for transcrip-

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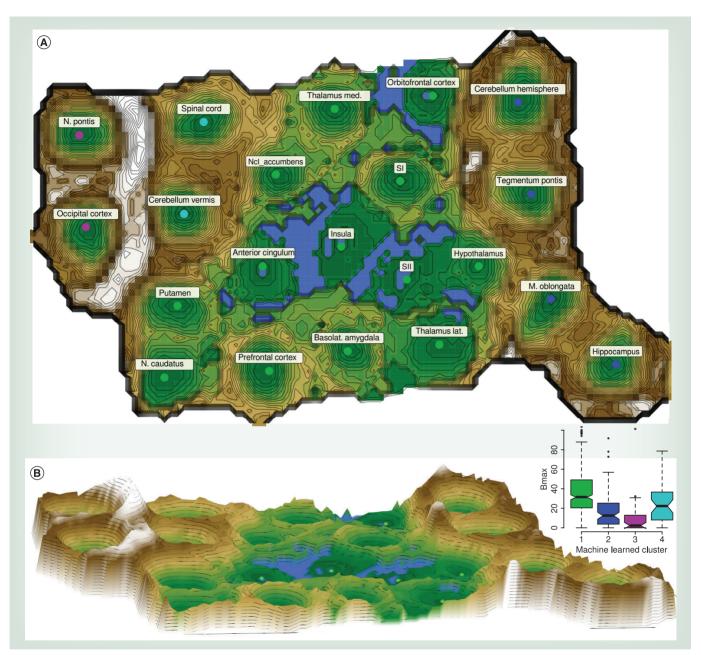


Figure 2. Pattern of μ -opioid receptor expression across 21 pain and addiction relevant areas of the human brain, analyzed from tissue acquired postmortem from 27 former heroin addicts. (A) Shows a projection of the log-transformed 21 × 27 B_{max} matrix onto a toroid neuronal grid where opposite edges are connected and using large numbers of nodes (4100). The dots indicate the so-called 'best matching units' of the self-organizing map, which are those neurons whose weight vector is most similar to the input. Neurons of the original 50 × 82 grid that do not match any input respectively brain regions are omitted. To the BMUs, the brain regions represented by them are annotated. The BMUs are colored according to the obtained classification respectively clustering of the data space. This cluster structure emerged from visualization of the distances between neurons in the high-dimensional space by means of a U-Matrix [22]. The U-Matrix was cut from a tiled display of the emergent self-organizing map to remove duplicate representation of the data. It was colored as a geographical map with brown or snow-covered heights and green valleys. Thus, valleys indicate clusters of similar brain regions. The watersheds of the U-Matrix indicate borderlines of clusters, which can be seen on the 3D-display of U-matrix shown below the (A) and which can be used as a haptic presentation of high-dimensional data by means of 3D-printing [43]. Between (A) and 3D views of the U*-matrix (B), the cluster differences with respect to the B_{max} values are shown as box and whisker plot constructed using the minimum, quartiles, median (solid horizontal line within the box) and maximum of the data, with the notches displaying a confidence interval around the median. The boxes are colored according to the BMUs cluster coloring in the self-organizing map.

BMU: Best matching unit.

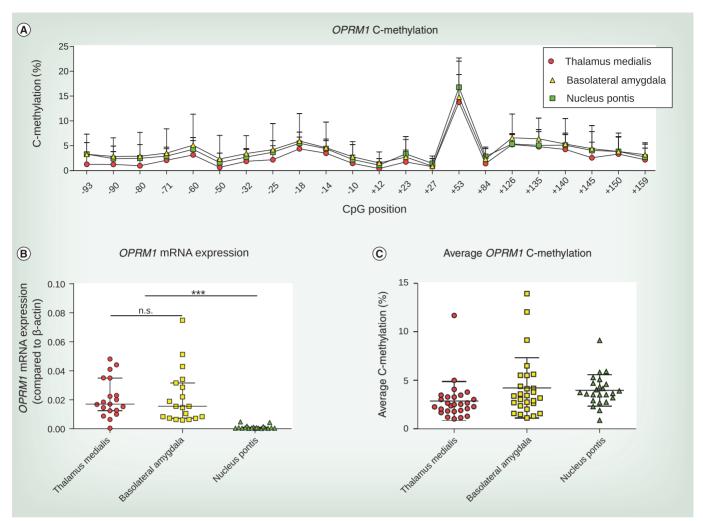


Figure 3. *OPRM1* C-methylation analysis and *OPRM1* mRNA expression in brain regions thalamus medialis, basolateral amygdala and nucleus pontis. (A) C-methylation profile across *OPRM1* gene from CpG position -93 to +159. Mean and SD from 24 to 27 brain tissue samples are shown. (B) *OPRM1* mRNA expression compared with β -actin differed in brain tissue of thalamus medialis, basolateral amygdala and nucleus pontis (Kruskal–Wallis $\chi^2 = 34.81$, df = 2, p < 2.8 × 10⁻⁸). Whereas thalamus medialis and basolateral amygdala showed no differences in *OPRM1* mRNA expression, both regions significantly showed significantly higher *OPRM1* mRNA expression levels than those found in nucleus pontis (Dunn's tests: ***p < 0.001, n.s.). Experiments were conducted from 19 brain tissue samples and median and interquartile ranges are shown. (C) Average *OPRM1* C-methylation of thalamus medialis, basolateral amygdala and nucleus pontis. Median and interquartile ranges are shown conducted from 24 to 27 brain tissue samples for each brain region. n.s.: Not significant.

tion. This further supports the possibility that DNA methylation is linked to the MeCP2-mediated chromatin structure of the *OPRM1* gene [4]. Further studies in different mice brain regions showed that μ -opioid receptor expression in microdissected mouse brain regions was negatively correlated with DNA methylation. In the cerebellum, which lacks μ -opioid receptors in the mouse brain, increased interaction with MeCP2 and a hypermethylated *OPRM1* promoter was observed. Moreover, a functional role of MeCP2 in association with the chromatin-remodeling factor Brg1 and with the DNA methyl transferase DNMT1 was established, suggesting that μ -opioid receptor expression can be epigenetically programmed in various mouse brain regions by MeCP2 [3]. Furthermore, the μ -opioid receptor upregulation observed during neuronal differentiation coincided with altered chromatin structure, an effect that occurred via recruitment of the transcriptional activator SP1 and by dissociation of repressors, such as MeCP2 and HDACs [5].

Assessments in postmortem brain tissue have been shown to be effective for the study of regulatory molecular mechanisms in the human brain [20,45]. While this approach allows direct investigations in the tissue of interest, the material raises several technical problems in contrast controlled *in vitro* assessments in cell cultures. The experimental conditions elude the strict control possible in cell cultures and are influenced by several

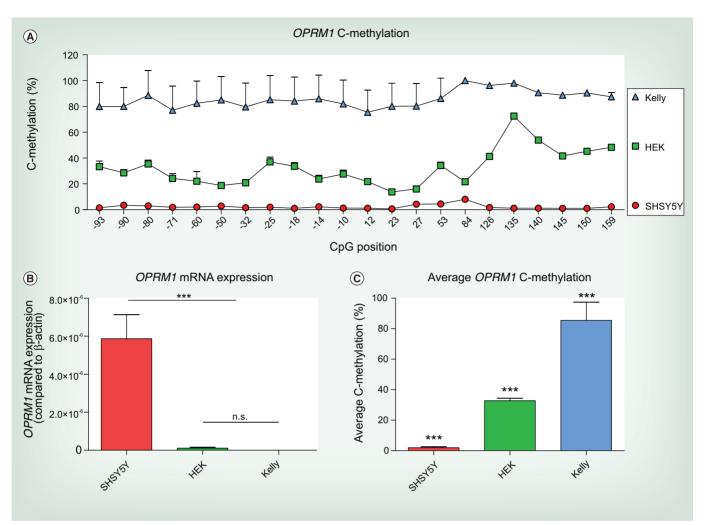


Figure 4. *OPRM1* C-methylation analysis and *OPRM1* mRNA expression in untreated SHSY5Y cells, HEK-293 cells and Kelly cells. (A) C-methylation profile across *OPRM1* gene from CpG position -93 to +159. Significant differences between different cell lines were assessed using rm-ANOVA with 'CpG position' as within-subject factor and cell type as between-subjects factor (main effect 'cell line' df = 2, F = 192.2, p = 2.04 × 10⁻¹¹ and 'CpG position' df = 20, F = 28.015, p = 1.4 × 10⁻⁵ and a significant interaction 'cell line' by 'CpG position' df = 40, F = 13.455, p = 9.5 × 10⁻⁵). (B) *OPRM1* mRNA expression of the untreated cell lines SHSY5Y, HEK-293 and Kelly. *OPRM1* mRNA expression is significantly higher in SHSY5Y cells compared with HEK-293 and Kelly cells, that do not express μ -opioid receptors (one-way ANOVA df = 2, F = 63.72, p < 0.0001; n.s. no significant difference between HEK-293 and Kelly cells, ***p < 0.001 for both HEK-293 and Kelly vs SHSY5Y). *OPRM1* mRNA expression compared with β -actin from three independent samples are shown as mean \pm SD. (C) Average *OPRM1* C-methylation of SHSY5Y, HEK-293 and Kelly cells. Significant differences between the cell lines were assessed using one-way ANOVA (df = 2, F = 216.7, p < 0.001) and *post hoc* t-test with Bonferroni α correction (all p < 0.001). Experiments were performed from six independent cell samples and are shown as mean \pm SD.

confounding factors arising before the subject's death. Processes occurring after death may also influence the quality of the biological material. DNA methylation, nevertheless, is a stable modification because of its covalent character and it is probably better conserved than are proteins and RNA. However, the collected brain material consists of approximately 70% glial cells (astrocytes, oligodendrocytes and microglia) [46] and therefore, the observed *OPRM1* methylation cannot be attributed solely to neurons. The consequences of this confounder, however, are probably ameliorated by the expression of opioid receptors in glial cells [47.48]. A further potential confounding factor is the previously reported influence of the *OPRM1* 118A>G polymorphism on μ -opioid receptor expression in the human brain [20]. Carriers of the variant G allele were also included in the present cohort and exploration of an effect of this variant on μ -opioid receptor expression was negative (details not shown). This agrees with our previous finding [18]. However, it contrasts with a later observation that the G allele reduces opioid receptor expression via a genetic–epigenetic interaction [20]. The latter result was only evident when a nonaddicted control group was available for comparison, which

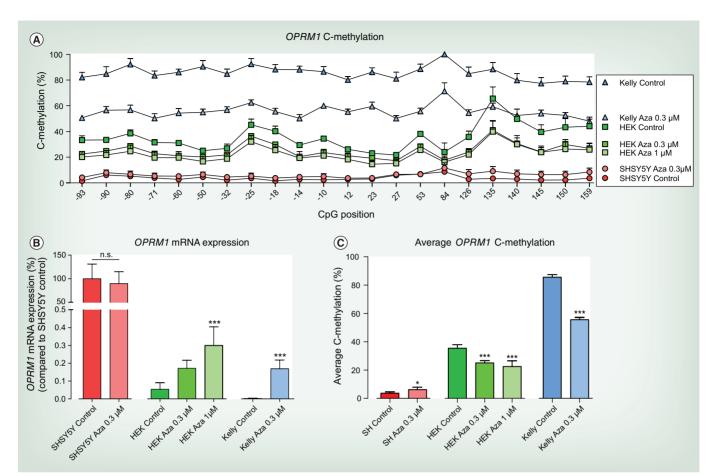


Figure 5. OPRM1 C-methylation analysis and OPRM1 mRNA expression in SHSY5Y cells, HEK-293 cells and Kelly cells treated with 5-Aza-2-desoxy-cytidine for 3 days. (A) C-methylation profile across OPRM1 gene from CpG position -93 to +159 of 5-Aza-2-desoxycytidine (5-Aza-CdR) treated and untreated SHSY5Y, HEK-293 and Kelly cells. Significant differences for each condition (cell line and treatment condition) were assessed using rm-ANOVA with 'CpG position' as within-subject factor and 'condition' as between-subjects factor (main effect 'condition' df = 6, F = 754.02, p = 1.32×10^{-29} and 'CpG position' df = 21, F = 58.123, p = 5.2×10^{-16} and interaction 'condition' by 'CpG position' df = 126, F = 18.416, p = 4.76×10^{-17}) and post hoc t-Test with Bonferroni α correction (Kelly Control vs Kelly Aza) 0.3 µM (p = 6.55 × 10⁻¹⁷), HEK-293 Control versus HEK-293 Aza 0.3 µM (p = 2.1 × 10⁻⁵) and versus HEK-293 Aza 1 µM (p = 6.6 × 10⁻⁹), SHSY5Y Control versus SHSY5Y Aza 0.3 μ M (n.s.)). (B) OPRM1 mRNA expression of 5-Aza-CdR treated and untreated SHSY5Y, HEK-293 and Kelly cells. OPRM1 mRNA expression is compared with SHSY5Y cells under control condition that were set as 100% to display different OPRM1 mRNA expression levels between cell lines. OPRM1 mRNA expression of 5-Aza-CdR treated HEK-293 and Kelly cells cannot reach OPRM1 expression levels of µ-opioid receptor positive SHSY5Y cells. To display the influence of 5-Aza-CdR treatment on OPRM1 mRNA expression for each cell line, 5-Aza-CdR treatment was compared with the respective control condition of the particular cell line. No significant differences were observed in OPRM1 mRNA expression in SHSY5Y cells following 5-Aza-CdR treatment (unpaired t-test p = 0.58). OPRM1 mRNA expression is significantly upregulated in HEK-293 cells following treatment with 1 µM 5-Aza-CdR (one-way ANOVA df = 2, F = 16.8, p = 0.0003; post hoc t-test ***p < 0.001). OPRM1 expression is significantly upregulated in Kelly cells following treatment with 0.3 µM 5-Aza-CdR (unpaired t-test ***p < 0.001). (A-C) At least three to six experiments were performed for each condition and are shown as mean plus SD. (C) Average OPRM1 C-methylation of 5-Aza-CdR treated and untreated SHSY5Y, HEK-293 and Kelly cells. Significant differences were assessed by comparing the 5-Aza-CdR treatment with the respective control condition of the particular cell line. Average OPRM1 methylation was significantly decreased following 5-Aza-CdR treatment in Kelly (unpaired t-test ***p < 0.001) and HEK-293 cells (one-way ANOVA df = 2, F = 28.17, p < 0.0001; post hoc t-test ***p < 0.001). In SHSY5Y cells 5-Aza-CdR treatment slightly increases the average OPRM1 methylation (unpaired t-test p = 0.0245) which was not significant when all CpG positions were considered (rm-ANOVA).

was not included in our mainly pharmacoepigenetic focused experiments.

A further limitation of the present analyses is the restriction to *OPRM1* DNA positions -93 to +159. When considering reports that the promoter spans positions from -445 to -10, with particularly important

promoter activity between positions -256 and -10 [26], the possibility exists that epigenetic regulation could have been found at other locations than those presently analyzed. The consequence of this limitation may be weakened when considering that a regulatory role of the presently analyzed *OPRM1* positions has been shown for a genetic-epigenetic interaction of the 118G allele [20]. Moreover, the analyzed region obviously sufficed to explain OPRM1 expression differences between the human cell lines. It should also be noted that the bisulphite-based pyrosequencing method used might have masked the occurrence of hydroxymethylcytosine as both modifications, in other words, hydroxymethyland methylcytosine, are recognized as DNA methylation. Hydroxymethylcytosine is markedly increased from the fetal to the adult brain with a strong enrichment in genic regions and distal regulatory elements. Furthermore, it occurs predominantly at the 5' splicing sites at exon-intron boundaries, suggesting novel roles in regulating splicing and gene expression. Hydroxymethylation is also an important modification of CpG sites at actively transcribed regions and may contribute to the epigenetic tuning at these regions [49].

The absence of differences in OPRM1 DNA methylation corresponding to differences in µ-opioid receptor expression in brain areas at the extreme ends of the opioid receptor expression range suggests that key regulatory mechanisms underlying these interregional differences target other mechanisms between gene transcription, post-transcriptional processes and protein assembly [50]. Due to the significant correlation between µ-opioid receptor protein and mRNA expression levels, our results narrow this window to posttranscriptional mechanisms acting before mRNA formation. The investigation of these mechanisms may require determinations during neuronal differentiation since the involvement of epigenetic mechanisms has previously been observed during neuronal differentiation of undifferentiated P19 cells [5]. In already differentiated cells, the gene locus is remodeled toward a transcriptionally active state, focusing the search on binding loci of transcription factors reported to be involved in the regulation of the human OPRM1 gene, such as activator protein 1 (AP1) [51], NF-κB [52], signal transducers and activators of transcription [53], GATA binding-protein (GATA) [54] and PARP [55], or several further negative transcription factors [56]. For example, different combinations of transcription factors could lead to a different expression pattern among brain regions, although validated 'cell specific' transcription factors have not been identified yet [50]. Gene expression is finally the result of the sum of several interacting epigenetic mechanisms, like histone modification, DNA methylation and miRNA that influence the transcriptional state of a gene. Epigenetic effects occurring concomitantly might mask other epigenetic mechanisms such as DNA methylation.

Conclusion

Following observations that exposure to opioids triggers epigenetic modifications [1,22-23,57], we analyzed whether such effects contribute to the differential expression of μ -opioid receptors across the human brain. Epigenetic regulation of the μ -opioid receptor via changes in DNA methylation was supported by several lines of evidence including *in vitro* assessments and analyses in brain tissue from rodents. Our present results support the possibility that *OPRM1* expression may be regulated via DNA methylation. However, they also indicate that in brain regions obtained postmortem from former opiate addicts, with a high degree of chronic opiate exposure, the changes in DNA methylation associated *in vitro* with a wide range of μ -opioid receptor expression levels greatly exceed the measured differences in *OPRM1* methylation.

Future perspective

Taking into account the fact that observations in laboratory animals could not be translated to the human setting, the field of epigenetic research still lacks evidence gathered from relevant human tissue [58]. However, the present study does not exclude that μ -opioid receptor expression is subject to epigenetic regulation also in humans. A broader analysis of all regulatory parts of the human *OPRM1* gene and a regard for other epigenetic mechanisms such as histone acetylation, which has also been shown to influence the μ -opioid receptor expression in rodents [44], needs still to be explored in human brain tissue to further refine the understanding of the mutual interactions between genetics, epigenetics and drugs in both, pain and addiction.

Supplementary data

To view the supplementary data that accompany this paper please visit the journal website at: www.futuremedicine.com/ doi/full/10.2217/epi-2016-0072

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Ethical conduct of research

The authors state that they have obtained appropriate institu-

tional review board approval or have followed the principles outlined in the Declaration of Helsinki for all human or animal experimental investigations. In addition, for investigations involving human subjects, informed consent has been obtained from the participants involved.

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Executive summary

- Accumulating evidence suggests that opioids may trigger DNA hypermethylation at global and possible single gen levels, in particular under chronic exposure. These epigenetic changes can be expected to alter the expression of genes.
- Of particular interest in this context is an effect of DNA methylation on μ-opioid receptor expression in the brain as the main site of clinical opioid action. A modulation of brain opioid receptor expression is supported by findings in rodents that have suggested a role of varying degrees of DNA methylation in the differential regulation of μ-opioid receptor expression across the brain.
- The role of μ-opioid receptor gene methylation for its expression in the human brain following chronic opioid exposure can be studied in postmortem acquired brain tissue from former heroin addicts. In the present study, this was done for 21 different brain regions selected for their involvement in pain, addiction or as control areas, sampled from 27 fatalities of Caucasian ethnicity.
- As the result of DNA methylation and μ-opioid receptor expression analysis, we observed that high versus low μ-opioid receptor expression, both observed at the functional proteomic (B_{max}) and the mRNA levels, were not associated with a similarly directed difference in the *OPRM1* methylation status.
- Subsequent *in vitro* experiments in human cell lines, however, indicated that different expression of μ-opioid receptors is indeed associated with similarly directed differences in *OPRM1* DNA methylation, however, the DNA methylation differed by a magnitude more than the difference observed between high and low *OPRM1* expression human brain areas.
- These results suggest that the pharmacoepigenetic effects of chronic opioid exposure are in principle possible but by their small effect sizes unlikely on brain µ-opioid receptor expression are unlikely to exert a major role in the region-specificity of µ-opioid receptor expression in the human brain.

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