Transcriptomic profiling of human hippocampal progenitor cells treated with antidepressants and its application in drug repositioning

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Abstract
Current pharmacological treatments for major depressive disorder (MDD) are ineffective in a significant proportion of patients, and the identification of new antidepressant compounds has been difficult. ‘Connectivity mapping’ is a method that can be used to identify drugs that elicit similar downstream effects on mRNA levels when compared to current treatments, and thus may point towards possible repositioning opportunities. We investigated genome-wide transcriptomic changes to human hippocampal progenitor cells treated with therapeutically relevant concentrations of a tricyclic antidepressant (nortriptyline) and a selective serotonin reuptake inhibitor (escitalopram). We identified mRNA changes common to both drugs to create an ‘antidepressant mRNA signature’. We used this signature to probe the Library of Integrated Network-based Cellular Signatures (LINCS) and to identify other compounds that elicit similar changes to mRNA in neural progenitor cells. Results from LINCS revealed that the tricyclic antidepressant clomipramine elicited mRNA changes most similar to our mRNA signature, and we identified W-7 and vorinostat as functionally relevant candidates, which may have repositioning potential. Our results are encouraging and represent the first attempt to use connectivity mapping for drug repositioning in MDD.

Keywords
Antidepressants, connectivity mapping, neural stem cells, major depression, drug repositioning

Introduction
Major depressive disorder (MDD) is a complex disorder characterised by a pathological distortion of affect (Jones et al., 2002). Antidepressants are the first line treatment for clinical MDD, but they are ineffective in a significant proportion of patients (Gibson et al., 2010). Evidence suggests that antidepressant drugs of different classes may evoke some convergent downstream effects at the messenger RNA (mRNA) level (Malki et al., 2012). Genetic factors can further moderate how initial upstream targets within each drug class relay their effects on to downstream transcriptional pathways (Malki et al., 2011; Powell et al., 2013). Consequently, increasing the number of available compounds with differing upstream mechanisms, but convergent downstream effects, will increase our chances of effectively treating the disorder in patients from different genetic backgrounds.

Identifying new types of antidepressant compounds, however, has been slow, with most being discovered through serendipity (Ban, 2006). Drawing from other areas of medicine, one way to identify new therapeutic compounds is through drug repositioning; the application of known drugs and compounds to new indications (Ashburn et al., 2004). There has been some promising evidence of drug repositioning in other fields of medicine using ‘connectivity mapping’, facilitated by the use of the Library of Integrated Network-based Cellular Signatures (LINCS; Lamb et al., 2006; Vempati et al., 2014). LINCS is essentially a mRNA library characterising the effects of over 20,000 small-molecule compounds, including over 1300 FDA approved drugs, in cell lines (Qu and Rajpal, 2012). LINCS allows users to identify compounds that elicit similar or opposite mRNA profiles to an experimentally-derived query mRNA signature (Qu and Rajpal, 2012).

Promising lines of research indicate that the hippocampus, and more specifically, hippocampal progenitor cells, are a key target of antidepressant medications (Boldrini et al., 2009; Malberg et al., 2000). Indeed, recent research suggests that antidepressant treatment in proliferating human hippocampal progenitor cells primes them for a route of advanced differentiation into neurons (antidepressant-induced hippocampal neurogenesis), which may be important in mediating their therapeutic effects (Anacker et al., 2011). Furthermore, in the rare cases where potentially new antidepressant compounds have been identified (e.g. the melatonergic analogue, agomelatine), they too seem to enhance...
neuroplasticity and neurogenesis in the hippocampus (Pompili et al., 2013).

This study attempts to tackle three main aims. First, we aim to investigate the effects of two antidepressants from different classes, a selective serotonin reuptake inhibitor (SSRI; escitalopram) and a tricyclic antidepressant (nortriptyline), on mRNA levels in human hippocampal progenitor cells. Second, we aim to isolate the overlapping effects of both drugs on gene expression, to create an ‘antidepressant mRNA signature’. Third we aim to use this antidepressant mRNA signature to probe the LINCS database and search for other compounds that elicit similar (and opposite) effects on mRNA to our signature, specifically in neural progenitor cells. The identification of known antidepressant compounds amongst those showing similar mRNA signatures to our query signature would confirm we have identified functionally meaningful mRNA targets. Furthermore, the identification of other functionally relevant compounds may point towards novel repositioning opportunities for MDD.

Methods

The hippocampal progenitor cell line

The multipotent, human hippocampal progenitor cell line HPC0A07/03C (provided by ReNeuron, Surrey, UK) was used for all experiments, as described previously (Anacker et al., 2011). ReNeuron’s HPC0A07/03C cells were obtained from a 12-week old foetus and immortalised with c-mycER technology. The hippocampal progenitor cell line represents a midrange between the treatment duration and the culture duration. Culture duration represents a midrange between the treatment duration and the culture duration. All cells were grown at 37°C, 5% CO2, and in a humidified atmosphere. Within these sets of experiments each ‘biological replicate’ represents a subculture of cells obtained from a different passage.

Drug doses

The active metabolite of the SSRI escitalopram, (S)-citalopram, has a therapeutic window of between 50–180 ng/mL in serum, which corresponds to doses of between 120–313 nM in vitro. The active metabolite of the tricyclic antidepressant nortriptyline, 10-hydroxy-nortriptyline, has a therapeutic window of between 70–170 ng/mL, which corresponds to 233–567 nM in vitro. Subsequently, cells were treated with a range of doses, incorporating two therapeutically-relevant doses of each drug and a high drug dose group comparable to those used previously in this cell line (Anacker et al., 2011).

Escitalopram drug doses were achieved by dissolving escitalopram oxalate (Sigma) in molecular grade ethanol (Sigma) to form a 10 mM stock solution. Escitalopram drug doses (0 nM, 145 nM, 290 nM and 1160 nM) were then formed by dilution of the stock with media, with the relative proportion of ethanol kept constant across all dose groups. Nortriptyline drug doses were achieved by dissolving nortriptyline hydrochloride (Sigma) in RNase free water to form a 10 mM stock solution. Nortriptyline drug doses (0 nM, 267 nM, 534 nM, 1068 nM) were then formed by dilution of the stock in media. To confirm that the drug doses were not toxic, we checked for any effects on cell death using immunohistochemistry. There was no evidence of increased cell death related to any dose in either drug condition (see Supplementary Information).

Culture protocol

Cells were seeded for 24 h on laminin-coated 6 well plates (Nunclon, Roskilde, Denmark). After seeding, cells were treated with media containing varying doses of escitalopram oxalate or nortriptyline hydrochloride for 48 h. At the end of the 48-h drug treatment period, proliferating media was aspirated and 1 ml of Trireagent (Sigma) was added for RNA isolation. Our 48-h drug culture duration represents a midrange between the treatment durations of neural progenitor cells in LINCS (6, 24 and 96 h).

Transcriptomics

RNA from cell experiments was isolated using Trireagent following the standard protocol, with an additional ethanol precipitation step to increase RNA purity. For gene expression experiments we utilised six biological replicates at four doses for each drug (total n = 24 per drug group). RNA samples were processed on Illuma Human HT-12 v4 Expression BeadChip (Illumina Inc., San Diego, CA) according to manufacturer’s protocol, see Supplementary Information.

Statistical analysis

Probes were normalised using a systematic approach incorporating the Lumi package (Du et al., 2008) and analysed using R (www.r-project.org). Probes were filtered to remove poorly detected probes and to investigate only the 7500 most variable probes (see Supplementary information online). We focused on the dose-dependent effects of each drug on gene expression. A linear regression was performed, where probe expression was selected as the dependent variable, dose as the independent variable, and batch and biological replicate included as factors. The false discovery rate (FDR) of multiple testing corrections was applied to our dataset with a q-value threshold of q < 0.05.

Gene ontology analysis

To understand which biological mechanisms may be affected by drug dose in a hypothesis-free manner we input p-values generated from our linear regressions from each of our drug groups into the Gene Ontology enRichment anaLysis and visualizAtion tool (GorillLa: http://cbl-gorilla.cs.technion.ac.il). All probes nominally affected by drug dose (p < 0.05) were tested for gene ontology (GO) term enrichment with biological processes (GOTERM_BP_FAT), cellular components (GOTERM_CC_FAT) and molecular function (GOTERM MF FAT). We utilised all probes present on the array after background correction, as our reference list.
To visualise gene co-expression networks relating to top GO terms and our antidepressant mRNA signature, we used the web-based application GeneMania (genemania.org). We input any nominally-significant transcripts identified in each condition into GeneMania, and selected the H. sapiens option, so that the tool would specifically draw connections between the genes in our experimentally-derived gene sets using human data only.

Gene co-expression networks

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Connectivity mapping

Connectivity mapping is the use of genome-wide expression data from cultured human cells treated with bioactive small molecules to discover functional associations between drugs, genes and diseases through the feature of shared gene expression changes (Lamb et al., 2006). The LINCS allows us to perform connectivity mapping and identify compounds which elicit similar downstream effects (positive connectivity scores) or opposite downstream effects (negative connectivity scores) on mRNA as our antidepressants. To implement connectivity mapping, we identified nominally-significant ($p < 0.05$) downregulated and upregulated transcripts in response to escitalopram dose or nortriptyline dose, or changes common to both drug doses, and input this information into LINCS (www.lincscloud.org). We then used the output to identify compounds (specifically) that elicit similar gene expression effects in neural progenitor cells (specifically). We considered any drugs as having a connectivity score of greater or equal to $\pm 0.75$ as potential candidates for repositioning.

Results

Escitalopram

Our results revealed no FDR-significant transcripts associated with escitalopram dose, but 324 were nominally (uncorrected $p < 0.05$) affected by dose. Our results revealed six FDR-significant ($q < 0.05$) GO terms in cells treated with escitalopram (Figure 1(a)); the top GO term was ‘extracellular exosome’, visualised as a gene co-expression network in Figure 1(b). We further input upregulated and downregulated transcripts associated with escitalopram dose into the LINCS database and identified eight compounds with the strongest positive connectivity scores (similar effects on mRNA) and eight compounds with the strongest negative connectivity scores (opposite effects on mRNA) in neural progenitor cells, visualised in Figure 1(c).
Our results revealed no FDR-significant transcripts associated with nortriptyline dose, but 458 were nominally (uncorrected \( p < 0.05 \)) affected by dose. Our results revealed nine FDR-significant (\( q < 0.05 \)) GO terms in cells treated with nortriptyline (Figure 2(a)); the top GO term was ‘intracellular organelle’, visualised as a gene co-expression network in Figure 2(b). We further input upregulated and downregulated transcripts associated with nortriptyline dose into the LINCS database and identified eight compounds with the strongest positive connectivity scores (similar effects on mRNA) and two compounds with the strongest negative connectivity scores (opposite effects on mRNA) in neural progenitor cells, visualised in Figure 2(c).

**Nortriptyline**

Our results revealed no FDR-significant transcripts associated with nortriptyline dose, but 458 were nominally (uncorrected \( p < 0.05 \)) affected by dose. Our results revealed nine FDR-significant (\( q < 0.05 \)) GO terms in cells treated with nortriptyline (Figure 2(a)); the top GO term was ‘intracellular organelle’, visualised as a gene co-expression network in Figure 2(b). We further input upregulated and downregulated transcripts associated with nortriptyline dose into the LINCS database and identified eight compounds with the strongest positive connectivity scores (similar effects on mRNA) and two compounds with the strongest negative connectivity scores (opposite effects on mRNA) in neural progenitor cells, visualised in Figure 2(c).

**Overlapping effects of antidepressants**

Thirty transcripts showed common nominally-significant changes in both drug groups (\( p < 0.05 \)); 20 out of 30 common transcripts were in the same direction in both drug groups (11 upregulated, 9 downregulated; Table 1), which formed our ‘antidepressant mRNA signature’. The overlapping transcripts are also visualised as a gene co-expression network (Figure 3(a)), and the eight compounds with the strongest positive connectivity scores (similar effects on mRNA), and eight compounds with the strongest negative connectivity scores (opposite effects on mRNA) in neural progenitor cells, are visualised in Figure 3(b).

**Discussion**

Our results revealed that two antidepressants with differing upstream mechanisms activate a set of both distinct and common downstream molecular pathways in hippocampal progenitor cells. Our GO analysis revealed that escitalopram specifically activated gene networks relating to extracellular organelles and vesicles, suggesting that it is modulating pathways related to extracellular communication (Figure 1). However, nortriptyline primarily affected intracellular components and organelles suggesting, conversely, that it modulates components of intracellular function (Figure 2).

We used expression changes associated with antidepressant dose to probe the LINCS database and to identify compounds which could be repositioned for the treatment of MDD. Neither antidepressant alone identified compounds with very high connectivity scores; all were below ±0.75 (Figures 1 and 2), so these compounds should be considered more cautiously than those with a higher connectivity score (e.g. those identified from our antidepressant mRNA signature (Figure 3) discussed below).
Identified compounds included those which were well-studied (e.g. loxapine), and relatively new compounds which may not have been tested in humans yet (e.g. BRD-K03354784), and so we will restrict our review of the suitability of compounds to those which are well-characterised.

In response to escitalopram dose, rutin was one of the top compounds with a positive connectivity score. Rutin is a citrus flavonoid found in a wide variety of plants such as citrus fruits (Davis, 1947). Numerous reports have revealed that rutin has antidepressant-like properties in mouse; for instance increasing mobility in the forced swim test (Machado et al., 2008; Nöldner and Schötz, 2002). Furthermore, it’s been identified as an essential antidepressant component to the medicinal herb, St John’s wort (*Hypericum perforatum*; Nöldner and Schötz, 2002). The compound with the highest positive connectivity score in relation to nortriptyline dose was loxapine. Loxapine is a typical antipsychotic medication used in the treatment of schizophrenia (Heel et al., 1978). Similarly to nortriptyline, loxapine has a rich pharmacology, and is metabolised by N-demethylation to amoxapine, a tetracyclic antidepressant (Cohen et al., 1982). Consequently, loxapine represents a compound with plausible repositioning potential for MDD.

Both escitalopram and nortriptyline are effective antidepressants and they also show common effects on mRNA levels that may represent a convergent mechanism (Table 1 and Figure 3). Overlapping transcripts include the zinc-finger transcription factor, insulinoma-associated 1 (*INSM1*), a gene which has previously been found to increase the generation and expansion of neural progenitor cells in the neocortex (Farkas et al., 2008). Therefore, this gene could be important in mediating increases in proliferation previously reported in response to antidepressants (Boldrini et al., 2012). Another gene also upregulated was oxidation resistance 1 (*OXR1*), which codes for a protein that specifically protects against oxidative stress (Volkert et al., 2000). This gene may be important in mediating protection against increases in reactive oxygen species hypothesised to trigger neuronal atrophy in the brains of MDD patients (Michel et al., 2012). Consequently, both antidepressants may be activating genes important in progenitor proliferation and neuroprotection, which may be important in mediating their effects.

Despite interesting compounds being identified when considering each drug separately (described above), the main purpose of our study was to identify common effects of escitalopram and

### Table 1. The transcripts affected by both escitalopram and nortriptyline dose, which formed our ‘antidepressant mRNA signature’.

<table>
<thead>
<tr>
<th>Direction of effect</th>
<th>Gene symbol</th>
<th>Gene name</th>
</tr>
</thead>
<tbody>
<tr>
<td>Upregulated</td>
<td>CASP6</td>
<td>Caspase-6</td>
</tr>
<tr>
<td></td>
<td>CENPQ</td>
<td>Centromere protein Q</td>
</tr>
<tr>
<td></td>
<td>EAPP</td>
<td>E2F-associated phosphoprotein</td>
</tr>
<tr>
<td></td>
<td>ID2</td>
<td>DNA-binding protein inhibitor ID-2</td>
</tr>
<tr>
<td></td>
<td>INSM1</td>
<td>Insulinoma-associated protein 1</td>
</tr>
<tr>
<td></td>
<td>MGC39900</td>
<td>(Hypothetical protein)</td>
</tr>
<tr>
<td></td>
<td>OSBP1B8</td>
<td>Oysterol-binding protein-related protein 8</td>
</tr>
<tr>
<td></td>
<td>OXR1</td>
<td>Oxidation resistance protein 1</td>
</tr>
<tr>
<td></td>
<td>SCGN</td>
<td>Secretagogin</td>
</tr>
<tr>
<td></td>
<td>SNX7</td>
<td>Sorting Nexin 7</td>
</tr>
<tr>
<td></td>
<td>SORBS2</td>
<td>Sorbin and SH3 domain-containing protein 2</td>
</tr>
<tr>
<td>Downregulated</td>
<td>CDC16</td>
<td>Cell division cycle protein 16 homologue</td>
</tr>
<tr>
<td></td>
<td>CSAD</td>
<td>Cysteine sulfenic acid decarboxylase</td>
</tr>
<tr>
<td></td>
<td>CXORF57</td>
<td>Chromosome X open reading frame 57</td>
</tr>
<tr>
<td></td>
<td>HS.25892</td>
<td>Small nucleolar RNA host gene 15 (non-protein coding)</td>
</tr>
<tr>
<td></td>
<td>KCNJ2</td>
<td>Potassium voltage-gated channel subfamily J member 2</td>
</tr>
<tr>
<td></td>
<td>LYAR</td>
<td>Cell growth-regulating nucleolar protein</td>
</tr>
<tr>
<td></td>
<td>MIR1974</td>
<td>microRNA 1974</td>
</tr>
<tr>
<td></td>
<td>SLC7A5</td>
<td>Solute carrier family 7 member 5</td>
</tr>
<tr>
<td></td>
<td>VGF</td>
<td>VGF nerve growth factor inducible</td>
</tr>
</tbody>
</table>

Figure 3. (a) Our antidepressant mRNA signature visualised as a gene co-expression network. Genes in grey represent hypothetical transcripts connecting other genes within the network, whereas genes in purple represent those transcripts identified from our dataset. (b) Eight compounds with the strongest positive connectivity scores (similar effects on mRNA) and eight compounds with the strongest negative connectivity scores (opposite effects on mRNA) in neural progenitor cells.
been shown to prevent the re-acidification of synaptic vesicles depressants (Dennis et al., 2011). Furthermore, bafilomycin has been shown to reverse some of the adverse cardiac effects of anti-mRNA signature (strongest negative connectivity score) was induce opposite downstream effects to antidepressants. The drug displaying the most potent reverse effects to our antidepressant mRNA signature (strongest negative connectivity score) was bafilomycin. Bafilomycin is an antibiotic which has previously been shown to reverse some of the adverse cardiac effects of antidepressants (Dennis et al., 2011). Furthermore, bafilomycin has been shown to prevent the re-acidification of synaptic vesicles in hippocampal tissue, potentially slowing the rate of neurotransmission across the synapse (Zhou et al., 2000). Therefore, functionally, bafilomycin may reverse some of the effects of antidepressants by slowing the release of neurotransmitters crossing the synapse (e.g. serotonin and noradrenaline), preventing the antidepressant’s initial mechanism of action, which is primarily to increase post-synaptic binding of serotonin and noradrenaline. However, this requires confirmation in further studies as the interactions between bafilomycin and antidepressants are not well understood, nor is it clear whether the effect of bafilomycin on synaptic vesicles is specific to the hippocampus.

Our results also revealed that the drug clomipramine had the highest connectivity score, suggesting it evokes similar changes to mRNA in neural progenitor cells as we identify in our study (Figure 3). Clomipramine is a tricyclic antidepressant used in the treatment of MDD with both serotonergic and noradrenergic effects (Amsterdam et al., 1997), and thus our results suggest that we may have identified a functionally valid set of mRNA changes, which can be used to identify compounds which elicit similar functional effects.

The drug ranked with the second highest connectivity score in the LINCS output was W-7 (N-(6-aminohexyl)-5-chloro-1-naphthalenesulfonamide hydrochloride). W-7 is an intracellular calcium binding protein (calmodulin) antagonist (Jan et al., 2000) and a functionally plausible compound for drug repositioning in MDD. Calmodulin antagonists and inhibition of calmodulin-dependent protein kinase II (CaMKII) disinhibit excitatory neurotransmission via presynaptic metabotropic glutamate receptors, and enhance neurotransmitter release (Hinds et al., 2003; O’Connor et al., 1999). Furthermore, antidepressants and calmodulin have both been shown to reverse corticosterone-induced gene transcription (Budziszewska et al., 2000); suggesting calmodulin inhibition reverses the downstream effects of stress, a known risk factor for MDD. Moreover, other studies have concluded that any interruption of the Ca2+-calmodulin-NOS-guanylyl cyclase subcellular signalling pathway will produce antidepressant-like effects (Paul, 2001).

The drug ranked with the third highest connectivity score is vorinostat, a histone deacetylase inhibitor, which has a similar mechanism of action to the mood-stabiliser sodium valproate (Kilgore et al., 2010). Both these medications are epigenetic modifiers which elicit potent effects on genome-wide gene expression (Covington et al., 2009). There is a considerable body of evidence which reveals epigenetic differences between MDD cases and controls (Saavedra et al., 2016). Furthermore, recent results from the genetic pathway analysis of an MDD genome-wide association study, suggest that genes regulating epigenetic modifications of histones might be causally related to MDD (The Network and Pathway Analysis Subgroup of the Psychiatric Genomics Consortium, 2015). Mechanistically, it could be that a genetic predisposition to MDD coincides with epigenetic modifications that result in reduced ‘transcriptional plasticity’, that impacts upon the propensity of hippocampal neural stem cells to differentiate (Burney et al., 2013). Indeed, hippocampal neural stem cell differentiation is associated with widespread transcriptional reprogramming (Mody et al., 2001), so a reduced propensity for transcriptional change may result in a reduced rate of neurogenesis, and therefore may explain the smaller hippocampal volumes observed amongst MDD patients (Bremner et al., 2000). Moreover, histone deacetylase inhibitors such as a vorinostat, that promote transcriptional activity across the genome, have previously been found to increase neurogenesis and neuroplasticity (Grayson et al., 2010). Thus, vorinostat may represent a functionally relevant compound with the potential to be repurposed in the treatment of MDD.

Despite the promising results, this study has a number of limitations. First, we investigated the acute effects of antidepressants on gene expression profiles. However, studies suggest that chronic treatment with antidepressants may be needed in order to observe therapeutic or functional effects (e.g. Malberg et al., 2000) and consequently we may be missing important gene expression changes linked to longer culture of progenitor cells with antidepressants. Thus, the short treatment duration may result in decreased ‘construct validity’ because it doesn’t reflect the same treatment duration it takes to observe clinical change, but because 48-h corresponds closely to the treatment duration in the LINCS database, the mRNA changes we identify may have good predictive validity in the context of drug screening (Powell et al., 2012).

Secondly, we cannot, using our methodology, accurately model the interactions of drugs in an in vivo environment where hormones and neurotransmitter release will moderate the effects of these drugs on neural stem cells. Furthermore, we can only provide indications as to what drugs might be promising repositioning opportunities for MDD; further functional work and, ultimately, clinical trials will be needed to confirm the repositioning validity of identified compounds.

In conclusion, this study is the first to report the genome-wide effects of two antidepressants on proliferating human hippocampal progenitor cells. We identify both overlapping and distinct effects of these drugs on mRNA levels. We isolated convergent effects of two antidepressants on 20 mRNA transcripts which we used to create an antidepressant mRNA signature to probe the LINCS database. Our results identified compounds which may reverse the effects of antidepressants (bafilomycin); compounds which elicit similar changes to mRNA, and are known to elicit the same functional effects (clomipramine); and compounds which show similar mRNA effects to our antidepressants with the potential to be repositioned as novel antidepressants (W-7, vorinostat). Our findings suggest that connectivity mapping and the use of the LINCS database may have relevance in the repositioning of compounds for the treatment of MDD.

Author’s Note

Sandrine Thuret and Gerome Breen contributed equally.
Declaration of conflicting interests

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References


