

# **Associations between regular cannabis use and brain resting-state functional connectivity in adolescents and adults**

Natalie Ertl<sup>1,2</sup>, Will Lawn<sup>3,4</sup>, Claire Mokrysz<sup>5</sup>, Tom P. Freeman<sup>5,7</sup>, Naji Alnagger<sup>5</sup>, Anna Borissova<sup>5,6</sup>, Natalia Fernandez-Vinson<sup>5</sup>, Rachel Lees<sup>7</sup>, Shelan Ofori<sup>5</sup>, Kat Petrilli<sup>7</sup>, Katie Trinci<sup>5</sup>, Essi Viding<sup>8</sup>, H. Valerie Curran<sup>5</sup>, Matthew B. Wall<sup>1,2,5\*</sup>.

<sup>1</sup>Invicro London, Burlington Danes Building, Hammersmith Hospital, Du Cane Road, London, UK

<sup>2</sup>Faculty of Medicine, Imperial College London, Hammersmith Hospital, Du Cane Road, London, UK

<sup>3</sup>Department of Psychology, Institute of Psychiatry Psychology and Neuroscience, King's College London, London, UK

<sup>4</sup>Department of Addictions, Institute of Psychiatry Psychology and Neuroscience, King's College London, London, UK

<sup>5</sup>Clinical Psychopharmacology Unit, University College London, 1-19 Torrington Place, London, , UK.

<sup>6</sup>Department of Neuroimaging, Institute of Psychiatry Psychology and Neuroscience, King's College London, UK

<sup>7</sup>Addiction and Mental Health Group (AIM), Department of Psychology, University of Bath, Bath, UK

<sup>8</sup>Clinical, Educational, and Health Psychology Research Department, University College London, London, UK

## **\*Corresponding Author:**

Matthew B. Wall

Burlington Danes Building

Hammersmith Hospital

Du Cane Road

London, W12 0NN, UK

Tel: 07801518566

[matt.wall@invicro.co.uk](mailto:matt.wall@invicro.co.uk)

**Running title: Associations between cannabis use and resting brain connectivity.**

**Keywords: Cannabis, fMRI, resting-state, striatum, CBD, THC, cannabinoids**

## **Abstract**

### Background/Aim

Cannabis use is highly prevalent in adolescents however little is known about its effects on adolescent brain function.

### Method

Resting-state functional Magnetic Resonance Imaging was used in matched groups of regular cannabis users (N=70, 35 adolescents 16-17 years old, 35 adults 26-29 years old) and non-regular-using controls (N=70, 35 adolescents/35 adults). Pre-registered analyses examined the connectivity of seven major cortical and sub-cortical brain networks (default mode network, executive control network, salience network, hippocampal network, and three striatal networks) using seed-based analysis methods with cross-sectional comparisons between user groups, and age groups.

### Results

The regular cannabis use group (across both age-groups), relative to controls, showed localised increases in connectivity only in the executive control network analysis. All networks showed localised connectivity differences based on age group, with the adolescents generally showing weaker connectivity than adults; consistent with developmental effects. Mean connectivity across entire network regions of interest (ROIs) was also significantly decreased in the executive control network in adolescents. However, there were no significant interactions found between age-group and user-group in any of the seed-based or ROI analyses. There were also no associations found between cannabis use frequency and any of the derived connectivity measures.

### Conclusion

Regular cannabis use is associated with changes to connectivity of the executive control network, which may reflect allostatic or compensatory changes in response to regular cannabis intoxication. However, these associations were not significantly different in adolescents compared to adults.

## Introduction

Cannabis is the most widely used illicit drug under international control, with 9.2% of 16-59 year olds reporting past year use in England and Wales (Office for National Statistics, 2022). This statistic increases to 18.6% in 16–24-year-olds. This pattern of high use in young people is reflected globally; data from 17 countries suggests the median onset of first use of cannabis is 18-19 years old (Degenhardt *et al.*, 2016). There is some recent evidence that cannabis use during adolescence may alter brain development in a number of ways (Jacobus and Tapert, 2014; Albaugh *et al.*, 2021).

A common method used to study human brain function is resting-state functional Magnetic Resonance Imaging (rs-fMRI). While the brain is at rest there are communities of structures which are highly functionally connected; resting-state networks (RSN). The most commonly studied RSN is the default mode network (DMN), which is most active when the brain is not actively engaged in a task (Raichle, 2015). The antagonistic network to the DMN, the executive control network (ECN), is the community of brain regions most active while engaged in an external task (Fox and Raichle, 2007; Seeley *et al.*, 2007). The Salience network (SN) is thought to be the mediator between these two networks as well as facilitating attention and detection of emotional and sensory stimuli (Goulden *et al.*, 2014). For further information on RSN see supplementary material.

As well as these cortical RSNs, other neural structures also show structured patterns of connectivity at rest. The functional connectivity of sub-cortical regions like the striatum and hippocampus can also be studied using fMRI (Wall, Freeman, *et al.*, 2022). The striatum can be sub-divided into three distinct regions based on functional and structural connectivity with the cortex: the limbic, associative, and sensorimotor divisions (Joel and Weiner, 2000) which are involved in motivational processes, cognition and motor functions respectively (Joel and Weiner, 2000; Martinez *et al.*, 2003).

Adolescence is a time of synaptic reorganisation and pruning in most mammals (Blakemore, 2008). During adolescence, there is a rapid development of the endocannabinoid system which contributes to maturation of corticolimbic neuron populations (Meyer, Lee and Gee, 2017). Resting-state fMRI studies have shown that during adolescence, neuronal networks become more segregated, leading to the hierarchical organisation seen in adulthood (Stevens, Pearlson and Calhoun, 2009).

Connectivity between frontal regions and the executive and salience networks increases with age and there is a migration of the DMN from central to more anterior and posterior positions (Solé-Padullés *et al.*, 2016). Animal literature has demonstrated CB<sub>1</sub>Rs are at their highest levels during adolescence, and decline into adulthood (Rodríguez de Fonseca *et al.*, 1993),

and dopamine receptors in the striatum are also over-produced prior to puberty and then heavily pruned during adolescence (Teicher, Andersen and Hostetter, 1995). Given that the main psychoactive component of cannabis is  $\Delta$ -9-tetrahydrocannabinol (THC), which is a partial agonist of the endogenous cannabinoid receptor 1 (Paronis *et al.*, 2012) (CB<sub>1</sub>R) there is cause for concern that regular use of cannabis in adolescents may alter corticolimbic development or other neurodevelopmental trajectories. One recent large study with a five-year follow-up period has shown a small association between adolescent cannabis use and cortical thinning in the medial prefrontal cortex (Albaugh *et al.*, 2021). Conversely, it has also recently been shown that there is no effect of age on reward anticipation and feedback processing in adolescent and adult cannabis use groups (Skumlien *et al.*, 2022).

There are mixed findings about the effects of cannabis on striatal function (Skumlien *et al.*, 2021), with lifetime cannabis use being previously associated with both increased (Nestor, Hester and Garavan, 2010) and decreased (Van Hell *et al.*, 2012) striatal activation when reward tasks are used. A selective effect of THC in the limbic striatum has recently been supported by work with an acute cannabis challenge, suggesting effects of THC on limbic striatum connectivity can be ameliorated when administered in combination with cannabidiol (Wall, Freeman, *et al.*, 2022). In other brain regions and networks, THC can cause morphological changes in the hippocampus (Chan, Hinds, Impey, & Storm, 1998), amygdala (Heath *et al.*, 1980) and cortex (Downer *et al.*, 2001), all of which highly express CB<sub>1</sub>R. Resting-state studies have shown decreased DMN connectivity in cannabis use groups compared to controls (Wetherill *et al.*, 2015; Ritchay *et al.*, 2021) while other studies have shown increases in functional connectivity in the core of the DMN but reduced connectivity with areas in overlapping networks (Pujol *et al.*, 2014). Further recent work has examined cerebellar-cortical connectivity (since the cerebellum also highly expresses the CB<sub>1</sub> receptor) and found relative increases in connectivity in the anterior cerebellum, with complementary decreases in the posterior regions (Schnakenberg Martin *et al.*, 2021). One study on incarcerated adolescents found cannabis use to be associated with decreased low frequency power in a number of networks including the DMN and the ECN. (Thijssen *et al.*, 2017). A recent meta-analysis of 21 cannabis-user resting-state fMRI studies (Thomson *et al.*, 2022) has identified a general pattern of increased connectivity in regular users, particularly in frontal-frontal, frontal-striatal, and frontal-temporal region pairings. This analysis also found an association between effects observed in these region pairings and the various measures of cannabis exposure used in the 21 studies (e.g., frequency of use, days of use per week, grams used per week). A more specific systematic review focussing on only studies in adolescents (Lichenstein *et al.*, 2022) found that resting-state connectivity of frontal, cingulate, and parietal regions was most often implicated in adolescent cannabis user groups, however also noted that the direction of these effects varies across studies.

No previous study has directly compared the effects of age (adolescents and young adults) and regular cannabis use vs controls on resting state-networks. To examine the differences in cortical and sub-cortical RSN connectivity in an adolescent versus adult regular cannabis use group and age matched controls as well as the interaction between cannabis usage and age, we used rs-fMRI on 140 subjects: 70 users (35 adults, 35 adolescents) and 70 controls (35 adults and 35 adolescents). The analysis method used was seed-based, or seed-to-voxel analysis. This type of functional connectivity analysis uses particular seed regions corresponding to major nodes of a network to define that network, for example, a region in the posterior cingulate cortex can be used to define the DMN. For further information on seed-based analysis please see the supplementary material. Three striatal networks (associative, limbic, and sensorimotor) and three cortical RSNs (DMN, ECN, SN), plus the hippocampal network, were defined in all subjects. We then used a similar seed-based approach to investigate significantly different user effect regions as well as a region of interest (ROI) approach to investigate global network differences, as pre-registered in our analysis plan on the Open Science Framework (Wall *et al.*, 2021). We hypothesised that connectivity will be reduced in the regular cannabis use group compared to controls. We also hypothesised that there will be an interaction between age-group and the cannabis use-group, such that adolescent cannabis use group will be more different to their age matched controls than the adult cannabis use group are to their age matched controls. Our final hypothesis was that cannabis use frequency will be positively correlated with measures of RSN connectivity in the cannabis use group.

## Methods

The data derives from the longitudinal arm of the 'CannTeen' study. Readers are directed to the full study protocol (Lawn *et al.*, 2019) for further specification of aims, data collection procedures, tasks, and power calculations for the full project. Other recent manuscripts report the full study and have focussed on cognitive effects and clinical symptoms in this cohort (Lawn *et al.*, 2022; Lawn *et al.*, 2022). Participants attended five behavioural sessions, one every three months, over the course of one year. Approximately half of the participants (see below) attended two MRI sessions, one at the start of the study, and the second a year later. The current data is a cross-sectional analysis of the baseline fMRI resting-state data. Our analysis plan was pre-registered (prior to any analysis taking place) here: <https://osf.io/jdvq7/> (Wall *et al.*, 2021).

## Ethical approval

Ethical approval was obtained from the University College London (UCL) Research Ethics Committee, project ID 5929/003. The study was conducted in line with the Declaration of Helsinki, and all participants provided written informed consent to participate.

## Participants

There were two between-subjects factors with two levels: cannabis use-group (regular cannabis use group and controls) and age-group (adolescents and adults). Participants were 70 current cannabis users and 70 age and gender-matched non-using controls, with an equal split of 35 adults and 35 adolescents in each group. In each of these subgroups there were 18 males and 17 females, except for the adult user-group which included 18 females and 17 males. Participants were recruited from the Greater London area via school assemblies, physical posters and flyers, and Facebook, Instagram, and Gumtree advertisements. Key inclusion criteria are displayed in Table 1. For a full list of all inclusion/exclusion criteria for the CannTeen study, please see the main study protocol.

	<b>Cannabis use group</b>		<b>Control group</b>	
	<b>Adolescent</b>	<b>Adults</b>	<b>Adolescent</b>	<b>Adults</b>
	(n=35)	(n=35)	(n=35)	(n=35)
<b>Inclusion criteria</b>	All participants Age 16-17 years or 26-29 years			
	<b>Cannabis use group</b>		<b>Control group</b>	
	Cannabis use at least one day per week, averaged over the past three months		Tobacco or cannabis use at least once in lifetime (*not required to be cannabis or tobacco naïve).	
<b>Exclusion criteria</b>	<b>All participants</b>			
	<ul style="list-style-type: none"> <li>• Personal history of a diagnosed psychotic episode or disorder</li> <li>• Current daily use of psychotropic medication • Past month treatment for any mental health condition, including cannabis dependence</li> <li>• BMI 34.9 (adults), or at 99.6th percentile (adolescents)</li> <li>• Any one illicit drug used more than two days per month, averaged over the past three months (except laughing gas)</li> <li>• Use of laughing gas more than once per week, averaged over the past three months</li> <li>• MRI contraindications</li> <li>• Left-handed</li> </ul>			
	<b>Cannabis use group (Adults only)</b>		<b>Control group</b>	
	Weekly or more frequent cannabis use before the age of 18, over a period of at least three months		More than 10 days of cannabis use in lifetime Any past-month cannabis use. More than one day of cannabis use in past three months	

Abbreviations: BMI = Body Mass Index, MRI = Magnetic Resonance Imaging

Table 1. Inclusion and exclusion criteria at baseline. \* Control group were not required to be cannabis and tobacco naïve for recruitment reasons.

### Data acquisition

Participants completed an MRI session shortly after their baseline behavioural session, at the Invicro clinical research facility, Hammersmith hospital, London, UK. The resting state



scan was eight minutes long and was acquired towards the beginning of the scanning session, after the anatomical scans, and a stop-signal task (reported elsewhere). Subjects were instructed to keep their eyes open but blink as normal for the duration of the scan, this was to prevent participants from falling asleep.

MRI data was collected with a 3.0 T Siemens Magnetom Verio scanner using a 32-channel phased array head coil. Standard shimming was applied throughout, and 'dummy' scans acquired before T1 stabilization had been reached were discarded automatically by the scanner. T<sub>2</sub>\* images were acquired using a multiband gradient echo Echo-Planar Imaging (EPI) sequence (TR = 1250 ms, echo time, TE = 30 ms, flip angle = 62°, parallel imaging factor = 2, multiband acceleration factor = 2, GRAPPA = 2, bandwidth = 1906Hz/pixel). A total of 384 volumes were collected for each subject, with a field-of-view of 192 mm and a matrix size of 64 x 64 mm, yielding an in-plane resolution of 3 x 3 mm. Slice thickness was also 3 mm, resulting in isotropic voxels. Forty-four slices were collected using an interleaved acquisition. Phase encoding direction was anterior to posterior. The forebrain, midbrain, and hindbrain (including the cerebellum) were covered. T<sub>1</sub>-weighted structural images were acquired using a Magnetization Prepared Rapid Gradient Echo (MPRAGE) sequence (TR = 2300 ms, TE = 2.98 ms, flip angle = 9°, parallel imaging acceleration factor = 2), with a spatial resolution of 1 mm isotropic.

## **Analysis**

All analysis procedures broadly followed the procedures for seed-based functional connectivity analyses used in previous independent datasets by (Comninos *et al.*, 2018; Demetriou *et al.*, 2018; Wall, *et al.*, 2019; Wall, Freeman, *et al.*, 2022).

### *Pre-processing*

Pre-processing and analyses of fMRI data was performed in FSL (FMRIB Software Library v6.0, Analysis Group, FMRIB, Oxford, UK), with the fMRI Expert Analysis Tool (FEAT; (Woolrich *et al.*, 2001; Smith *et al.*, 2004))

Structural high resolution (anatomical) images were pre-processed using the `fsl_anat` function, which implements brain extraction, bias field correction, normalisation, and tissue segmentation. Pre-processing of the functional data consisted of head motion correction (with MCFLIRT), brain extraction (with BET) temporal filtering (100s), and spatial smoothing (6mm FWHM Gaussian kernel). The functional images were then normalised to MNI-152 (Montreal Neurological Institute) space with FNIRT (FMRIB's non-linear registration tool), using a 10 mm warp resolution and 12 degrees of freedom.

In order to ensure the quality of the data, each subject’s raw functional image series was inspected for severe motion (>3 mm max displacement) and other artifacts. Outcomes of the registration process were also considered. Two subjects were excluded for excessive head motion after these checks leaving the final sample of N = 138. One adult control and one adolescent control were excluded at this stage.

*First-level analysis*

First-level analyses used seed-based functional connectivity methods, where a time-series from a particular region (the ‘seed’) is used to interrogate data from the rest of the brain in order to identify other areas that have correlated time-series; the implication being that areas with similar temporal characteristics are functionally connected. See Table 3 for a list of approximate centre of gravity coordinates for each seed. The regions for the PCC and Anterior Insula seeds were the same as those used in (Wall, *et al.*, 2019). These were derived from automated meta-analytic data on <http://neurosynth.org/> using the ‘default mode’ and ‘salience’ terms (uniformity tests). In a divergence from our pre-registered analysis plan, we used an additional region in the dorsolateral prefrontal cortex (DLPFC; as recommended by (Yeo *et al.*, 2011)) as the seed-region for the executive control network, after testing showed this gave a better definition for the ECN that was more similar to previous work. This region was also derived from <http://neurosynth.org/>, using the “executive control” term. These meta-analysis maps were thresholded at an appropriate level (Z = 12, 10, and 6 for the default mode, salience, and executive control maps, respectively) to achieve anatomically plausible regions, and the PCC, DLPFC, and anterior insula clusters isolated and binarized for use as image masks. The hippocampus seed region, denoting the hippocampal network, was defined anatomically using the Harvard-Oxford subcortical atlas (see supplementary figure 1A).

Masks for the three striatal networks (associative, limbic, and sensorimotor) were the same as those used in (Wall, Freeman, *et al.*, 2022) and are defined according to the original parcellation by Martinez (Martinez *et al.*, 2003) and using the atlas provided by Tziortzi *et al.* (2013). The associative mask includes the precommissural dorsal caudate, the precommissural dorsal putamen, and postcommissural caudate. The limbic mask includes the ventral caudate and substantia nigra, and the sensorimotor mask comprises the postcommissural putamen (see supplementary figure 1B).

<b>Seed Region</b>	<b>X</b>	<b>Y</b>	<b>Z</b>
--------------------	----------	----------	----------

Anterior Insula	(R) 38.90	(R) 22.33	(R) -3.38
	(L) -36.24	(L) 20.14	(L) -3.84
Associative striatum	(R) 23.88	(R) 7.93	(R) 2.07
	(L) -23.88	(L) 7.93	(L) 2.07
Limbic striatum	(R) 16.07	(R) 10.88	(R) -8.39
	(L) -16.62	(L) 11.97	(L) -8.39
Sensorimotor striatum	(R) -29.15	(R) -7.64	(R) 2.98
	(L) -29.15	(L) -7.64	(L) 2.98
Dorsolateral prefrontal cortex	(R) 47.54	(R) 8.84	(R) 28.91
	(L) -46.63	(L) 13.84	(L) 29.37
Hippocampus	(R) 28.06	(R) -18.54	(R) -16.12
	(L) -25.34	(L) -17.45	(L) -17.03
Posterior Cingulate Cortex	-0.22	-51.67	27.09

Table 2: Approximate centre of gravity coordinates in MNI152 standard space of the seven seed regions.

All these standard (MNI152) space mask images were coregistered to each individual subject's functional space, thresholded at 0.5, and binarised to produce the final individualised mask images. The mean time series from each mask was extracted from the functional image series for each participant's scan. These time-series were then used as regressors in individual (one per region) first-level analysis models. Each of these regressors of interest were used in separate first level models for each participant. This produces individual maps of functional connectivity for each participant for each functional network, defined by their relationship to activity in the seed region.

Mean white matter (WM) and cerebro-spinal fluid (CSF) masks were also produced as part of the anatomical image processing, by FSL's FAST algorithm. These masks were also co-registered to each subject's functional space and thresholded at 0.5. Mean signals from these masks were extracted and included in each model as regressors of no interest to reduce the effect of noise, along with an extended set of motion parameters including temporal derivatives and quadratic versions of the six (three translations, and three rotations) basic motion parameters. The inclusion of WM and CSF regressors is similar to the CompCor approach (Behzadi *et al.*, 2007), is a principled, robust, and effective method of reducing the influence of a range of noise sources (e.g. physiological, motion, thermal), and is useful for both resting-state (Comninos *et al.*, 2018; Demetriou *et al.*, 2018; Wall, Lam, *et al.*, 2022) and task fMRI data (Thurston *et al.*, 2022). Functional connectivity with

each seed region was assessed using a positive contrast for the regressor of interest (the time series from each seed-region) in each model.

### *Second level analysis*

All second level analyses used FMRIB's local analysis of mixed effects (FLAME); a two-step process using Bayesian modelling and estimation. FLAME uses a weighted least-squares approach and does not assume equal variance between groups. All group-level analyses used cluster-level thresholding (Friston *et al.*, 1994; Woo, Krishnan and Wager, 2014) with a cluster-defining threshold of  $Z = 2.3$  and a multiple-comparisons corrected cluster-extent threshold of  $p < 0.05$ , in order to account for multiple comparisons.

As an initial validation of the methods and analysis approach, a simple mean group-level analysis was computed for all seven seed-regions. This analysis collapsed across all subjects and age/cannabis use group, and the results (see supplementary figures 2 and 3) were compared to derived networks in previous similar work (Wall, *et al.*, 2019; Wall, Freeman, *et al.*, 2022) as a validation step. Next, between-group analyses were conducted on the seed-to-voxel data using a fully factorial two-way between-subjects ANOVA design to test the effects of age-group, cannabis use-group, and their interaction. F tests were used in the contrasts of this model to reveal significant differences between groups, and significant interaction effects. Analyses of this type produce maps of F statistics which (unlike t statistics used for simple contrasts) are non-directional (always positive) and are therefore uninformative as to the direction of the effects. Therefore, the significant clusters resulting from these ANOVA analyses were defined as ROIs, and mean values were extracted from these regions for each participant. These values were then plotted to determine the precise pattern and direction of the effects across the four groups.

### *Network ROI analyses*

The mean network maps produced by the initial group-mean validation analyses were also used to define a broad ROI for each functional network, facilitating additional analyses. Maps were thresholded at  $Z = 80\%$  of maximum to define a plausibly anatomically-constrained set of regions, these are outlined in table 3. These were then binarized to produce network masks, which data could then be extracted from to give estimates of network differences in the groups.

---

<b>Network</b>	<b>Threshold 80% Z level</b>
----------------	------------------------------

Associative striatum	9.55
Limbic striatum	7.88
Sensorimotor striatum	8.29
DMN	11.80
ECN	10.13
Saliency Network	10.97
Hippocampus	9.06

Table 3. Z threshold levels used to define network regions of interest, defined at 80% of the maximum Z value in that group network.

Parameter estimates (single values, representing overall connectivity within the network) were then extracted for each subject, and for each network (as in (Wall, *et al.*, 2019; Wall, Freeman, *et al.*, 2022)). These individual-level connectivity measures were then analysed using ANOVA to test for main effects of age, cannabis use status, and an interaction in the overall network-level results. These extracted parameter estimates from each resting-state network were also correlated with cannabis use frequency in the cannabis using groups (separate correlation analyses for each age group), in order to address our final hypothesis. The alpha threshold for these correlations was reduced to 0.007 to reflect the seven tests (across the seven networks) being conducted. These analyses were all conducted using Jamovi version 2.3.21.0.

## Results

### *Participants*

A summary of the participant characteristics (demographics, questionnaire scores, and drug history) can be found in Table 1. Maternal education was recorded as a socioeconomic marker. Beck Depression Inventory (BDI) and Risk-Taking-18 (RT-18) were recorded to measure other factors which may affect drug taking behaviour and control for potential differences across the four groups as closely as possible.

	Adolescent cannabis use group	Adult cannabis use group	Adolescent controls	Adult controls	Group differences
--	-------------------------------------	--------------------------------	------------------------	-------------------	-------------------

	(n = 35)	(n = 35)	(n = 34)	(n = 34)	
<b>Gender, n (%)</b>					<i>Age difference</i>
Female	17 (49%)	18 (51%)	17 (50%)	16 (47%)	$\chi^2(1, N = 138)=0.00, P=1$
Male	18 (51%)	17 (49%)	17 (50%)	18 (53%)	<i>User difference</i> $\chi^2(1, N = 138)=0.03, P=0.86$
<b>Age in years, mean (SD)</b>	17.23 (0.50)	27.88 (1.12)	17.19(0.46)	27.47 (0.94)	*** <i>Age main effect Adult&gt;Adolescent</i> F(1,134)=5781, P<0.001
					<i>User main effect</i> F(1,134)=2.20, P=0.14
					<i>Interaction</i> F(1,134)=2.12, P=0.15
<b>Maternal education, n (%)</b>					<i>Age difference</i>
Below undergraduate degree	15 (43%)	15 (43%)	13 (38%)	20 (59%)	$\chi^2(1, N = 136)=1.45, P=0.23$
Undergraduate degree or above	19 (54%)	20 (57%)	21 (62%)	13 (38%)	<i>User difference</i> $\chi^2(1, N = 136)=0.456, P=0.49$
<b>Education to the highest-level mean (SD)</b>					
<i>Adults only</i>		5.46 (1.67)		6.24 (1.37)	* <i>User difference</i> t[67]=2.03 P=0.046
1 = secondary education only, 8 = doctoral-level qualification					
<b>BDI, mean (SD)</b>	10.57 (6.32)	8.46 (10.47)	9.50 (6.31)	7.62 (8.52)	<i>Age main effect</i> F(1,134)=2.09, P=0.15
					<i>User main effect</i> F(1,134)=0.47, P=0.49
					<i>Interaction</i> F(1,134)=0.00, P=0.93
<b>RT-18, mean (SD)</b>	11.66 (3.38)	7.86 (4.03)	8.53 (3.99)	7.41 (4.59)	*** <i>Age main effect Adolescent&gt;Adult</i> F(1,134)=12.94, P<0.001
					** <i>User main effect</i> <i>User&gt;Control</i> F(1,134)=6.83, P=0.01
					<i>Interaction</i> F(1,134)=3.85, P=0.052
<b>Alcohol use, days/week, mean (SD) [min-max]</b>	0.86 (0.76) [0-3.25]	1.55 (1.33) [0-5.25]	0.8 (0.85) [0-3.67]	1.45 (1.24) [0-5.25]	*** <i>Age main effect Adult&gt;Adolescent</i> F(1,134)=13.53, P<0.001
					<i>User main effect</i> F(1,134)=0.17, P=0.69
					<i>Interaction</i> F(1,134)=0.01, P=0.90
<b>Cigarette/roll-up use, days/week, mean (SD) [min-max]</b>	2.19 (2.74) [0-7]	1.64 (2.77) [0-7]	0.67 (1.82) [0-6.58]	0.63 (1.8) [0-7]	<i>Age main effect</i> F(1,134)=0.19, P=0.66
					<i>User main effect</i> F(1,134)=0.00, P=0.93
					<i>Interaction</i>

<b>Other illicit drug use, monthly use, n (%)</b>					
Yes	20 (57%)	8 (23%)	1 (3%)	0 (0%)	** Age main effect Adult>Adolescent $\chi^2(1, N = 138)=7.38, P=0.007$ *** User main effect User>Control $\chi^2(1, N = 138)=30.8, P<0.001$
No	15 (43%)	27 (77%)	33 (97%)	34 (100%)	
<b>Cannabis Use Ever use (controls), n (%)</b>			30 (88%)	33 (97%)	Age difference t[66]=1.39, P=0.17
<b>Number of lifetime uses</b>	1193.13	271.52			*** Age effect F[1,133]=24.6, P<0.001
mean (SD) [min-max] (rough estimate in users*)	(1032.56) [20.77 - 4173.75]	(248.12) [14.89 - 1102.94]	3.59 (2.95) [0-10]	4.65 (3.18) [0-10]	*** User effect F[1,133]=61.3, P<0.001 *** Interaction F[1,133]=24.5, P<0.001
<b>Days/week of use (users), mean (SD) [min-max]</b>	3.35 (2.13) [0.83-6.92]	3.83 (2.10) [0.75-6.92]			Age difference t[68]=0.938, P=0.35
<b>Grams used on a day of use (users), mean (SD) [min-max]<sup>b</sup></b>	0.97 (0.81) [0.15-4]	0.72 (0.84) [0.03-3.5]			Age difference t[67]=1.25, P=0.21
<b>Hours since last use (users), mean (SD) [min-max]<sup>c</sup></b>	44.07 (31.79) [12.5 - 136]	42.10 (42.73) [12.08-185]			Age difference t[66]=0.41, P=0.68
<b>Age of first ever use (users), mean (SD)</b>	14.72 (1.09)	18.05 (3.27)			*** Adult>Adolescent Mann-Whitney U = 228, P<0.001
<b>Age of first weekly use (users), mean (SD)</b>	15.77 (1.03)	22.29 (2.95)			*** Adult>Adolescent t[68]=12.4, P<0.001
<b>CUDIT (users), mean (SD)</b>	15.37 (6)	11.86 (5.7)			**Adolescent>Adult t[68]=2.51, P=0.01
<b>DSM-5 severe CUD (users), n (%)</b>	16 (46%)	7 (20%)			* Adolescent>Adult t[68]=2.35, P=0.02

Table 4. Demographic, questionnaire, and drug history information of the participant sample. Sociodemographic characteristics of full sample minus the two subjects excluded for head motion (final n=138). BDI is Beck Depression Inventory. RT-18 is Risk-Taking-18. CUDIT is the cannabis use disorder inventory test. Continuous data are presented as mean [SD], and categorical data are presented as n (%). \* Rough estimate of life-time cannabis use in the cannabis use group was calculated by

multiplying total years of use (age – age first weekly use) and uses per year (weekly use x 52). Group differences are highlighted in the final column using appropriate tests for each data type ( $\chi^2$ , ANOVA, Mann-Whitney U, and t-tests; \*P<0.05, \*\*P<0.01, \*\*\*P<0.001).

### Head motion

To examine whether head motion differed between the groups, an ANOVA was conducted on the absolute mean displacement values (mm) derived from the head-motion correction process. We found no significant effects: of cannabis use:  $F[1,136]=0.318$ ,  $P=0.573$ , of age:  $F[1,136]=0.20$ ,  $P=0.656$ , or interaction:  $F[1,136]=0.108$ ,  $P=0.743$ .

### User-group main effects: Seed-to-voxel analyses

Details of significant clusters for ‘cannabis use effects’ can be found in Supplementary Table 2. Regions were labelled using a mixture of Harvard-Oxford cortical and subcortical atlases (Desikan *et al.*, 2006) and expert knowledge of functional areas. User group main effects were only found in the executive control network analysis, with no other significant clusters in any other network. The ECN results are shown in figure 1 and show significant clusters in the dorsal anterior cingulate cortex which extends into a cluster in the motor cortex, the middle and posterior insula, the temporo-parietal junction (TPJ) and the superior temporal gyrus.

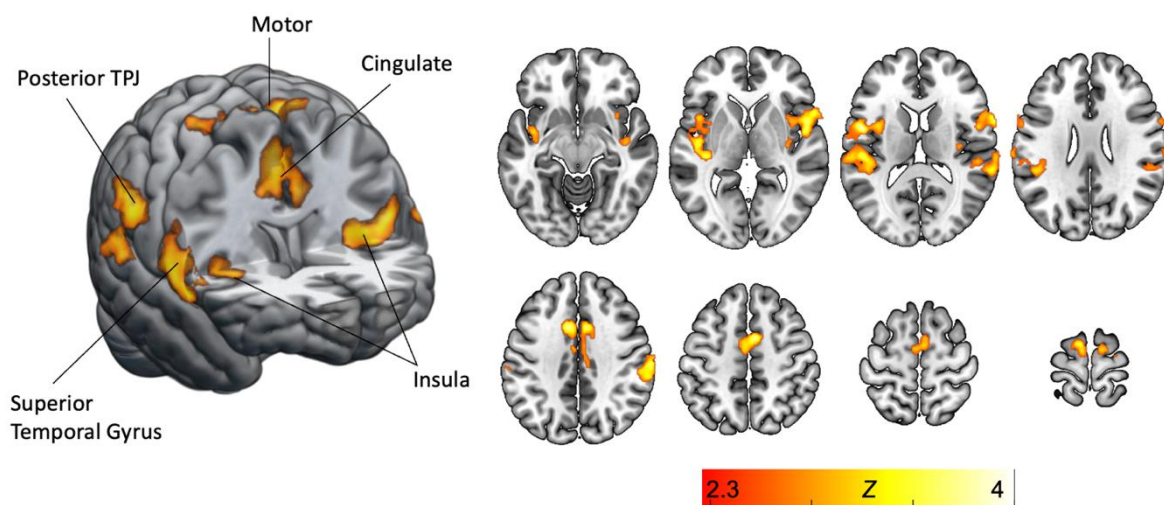


Figure 1: Areas of significant differential connectivity with the executive control network (dorsolateral prefrontal seed) between the cannabis use group and the



control group. Background image is the MNI152 standard template brain. Images in neurological format (left of the image = left hemisphere).

The pattern of connectivity seen in the user-group effects showed a slight migration of functional hubs from the canonical ECN. Figure 2 shows the user-group differential connectivity result overlaid with the mask derived from the group mean result (in blue; see supplementary material).

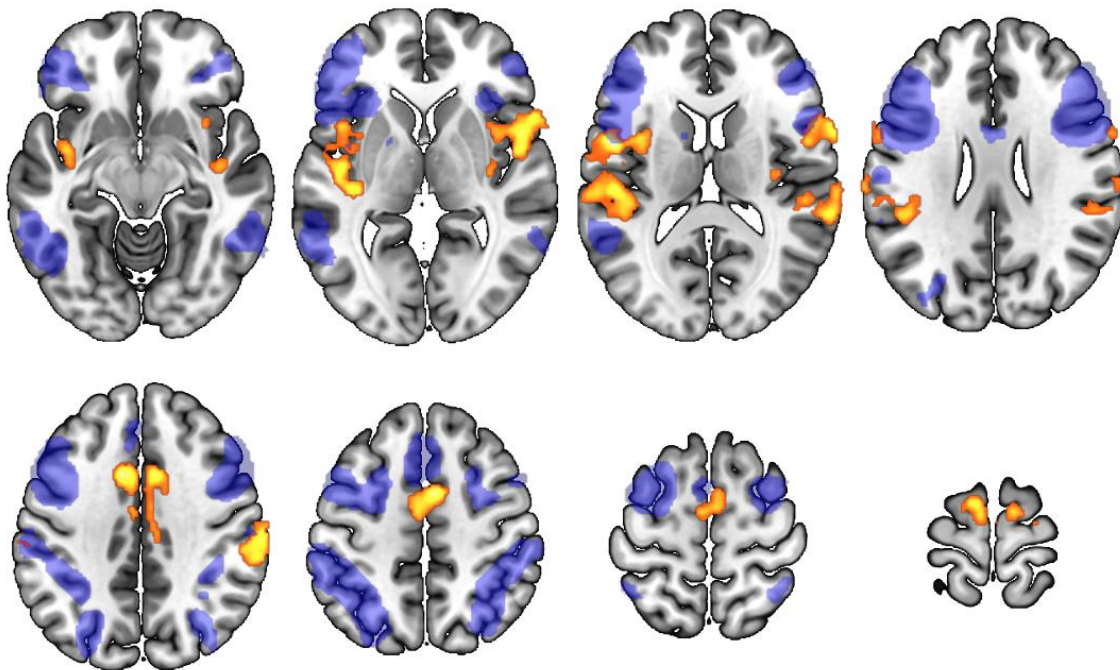


Figure 2: Executive control network (ECN) user-group effects (yellow-orange) overlaid with the group average (blue) ECN connectivity (thresholded at  $Z = 10.13$ ). Background image is the MNI152 standard template brain. Images in neurological format (left of the image = left hemisphere).

To identify the pattern of directionality of these results, clusters were split into ROIs and  $2 \times 2$  (user group  $\times$  age group) between-subjects ANOVAs were performed. These results are visualised in figure 3. All areas showed greater connectivity in the cannabis use groups compared to control groups. Posterior Temporo-parietal Junction (TPJ):  $F(1, 134) = 16.96$ ,  $P < 0.0001$ , Motor:  $F(1, 134) = 18.52$ ,  $P < 0.0001$ , Cingulate:  $F(1, 134) = 21.30$ ,  $P < 0.0001$ ,

Superior Temporal Gyrus:  $F(1, 134) = 18.43$ ,  $P < 0.0001$ , Insula:  $F(1, 134) = 21.49$ ,  $P < 0.0001$ ,  $N=138$ .

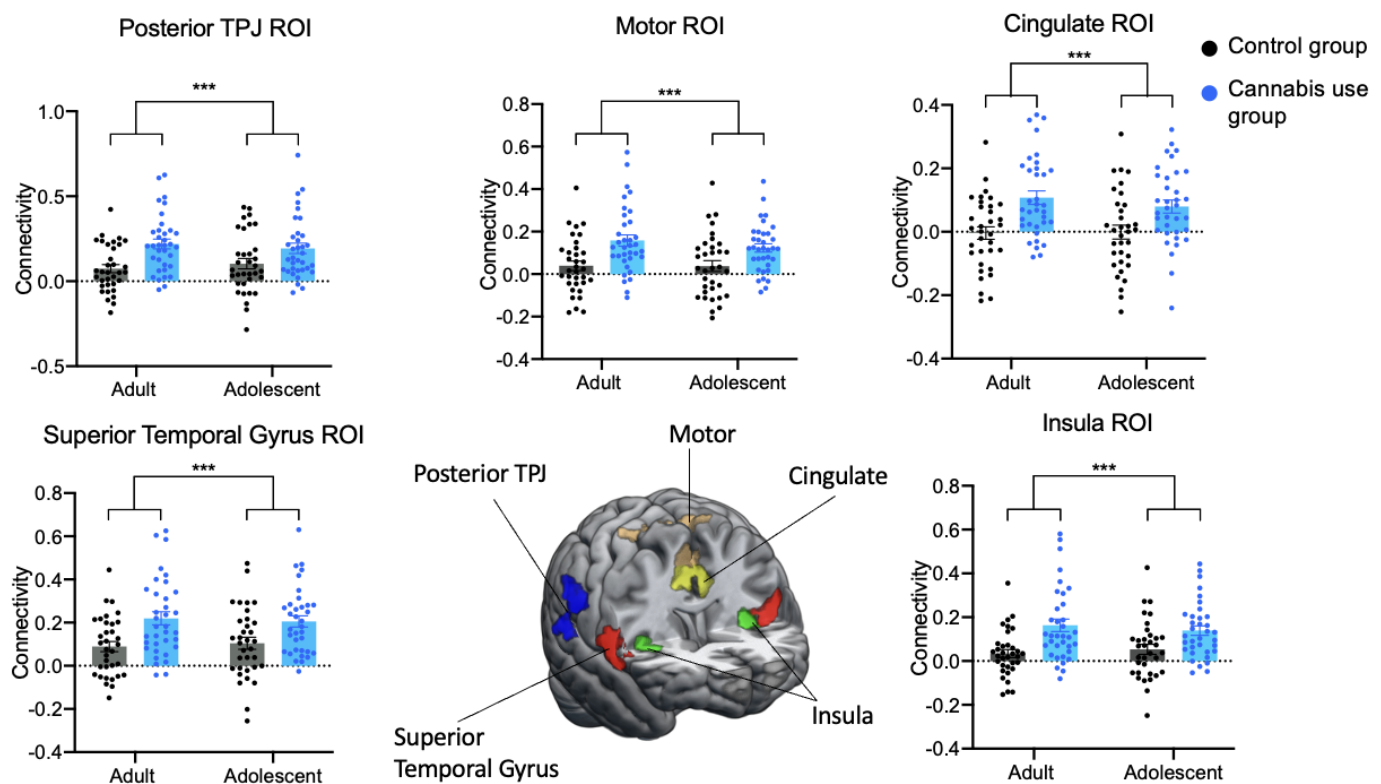


Figure 3: Areas of significantly different connectivity (found from seed-voxel analysis) in the executive control network (dorsolateral prefrontal seed) between the cannabis use groups and the control groups, divided into five regions of interest to assess directionality using histograms and between-subjects ANOVAs,  $N=138$  \*\*\* $P < 0.001$ .

#### Age-group main effects: Seed-to-voxel analyses

Age group main effect clusters were found in all seven networks, these are summarized in figure 4. Follow-up analyses revealed that connectivity was generally lower in the adolescent group compared to the adult group, with the exception of the salience network where the regions were relatively more connected in the adolescent groups. Statistical outliers were removed (one data point from the DMN, three from the limbic striatum and one from the sensorimotor striatum) using the ROUT method ( $Q=1\%$ ); however, this did not alter the overall significance. ECN:  $F(1,134) = 56.98$ ,  $P < 0.0001$ ,  $N138$ , Associative striatum:  $F(1,134) = 37.41$ ,  $P < 0.0001$ ,  $N138$ , DMN:  $F(1,133) = 27.20$ ,  $P < 0.0001$ ,  $N137$ , Salience network:  $F(1,134) = 53.62$ ,  $P < 0.0001$ ,  $N138$ , Hippocampal network:  $F(1,134) = 20.03$ ,  $P < 0.0001$ ,

N138, Limbic striatum:  $F(1,131) = 40.82, P < 0.0001, N135$ , Sensorimotor striatum:  $F(1,133) = 52.21, P < 0.0001, N137$ . Summary histograms can be found in supplementary figure 5 and the derived ROIs in supplementary figure 6.

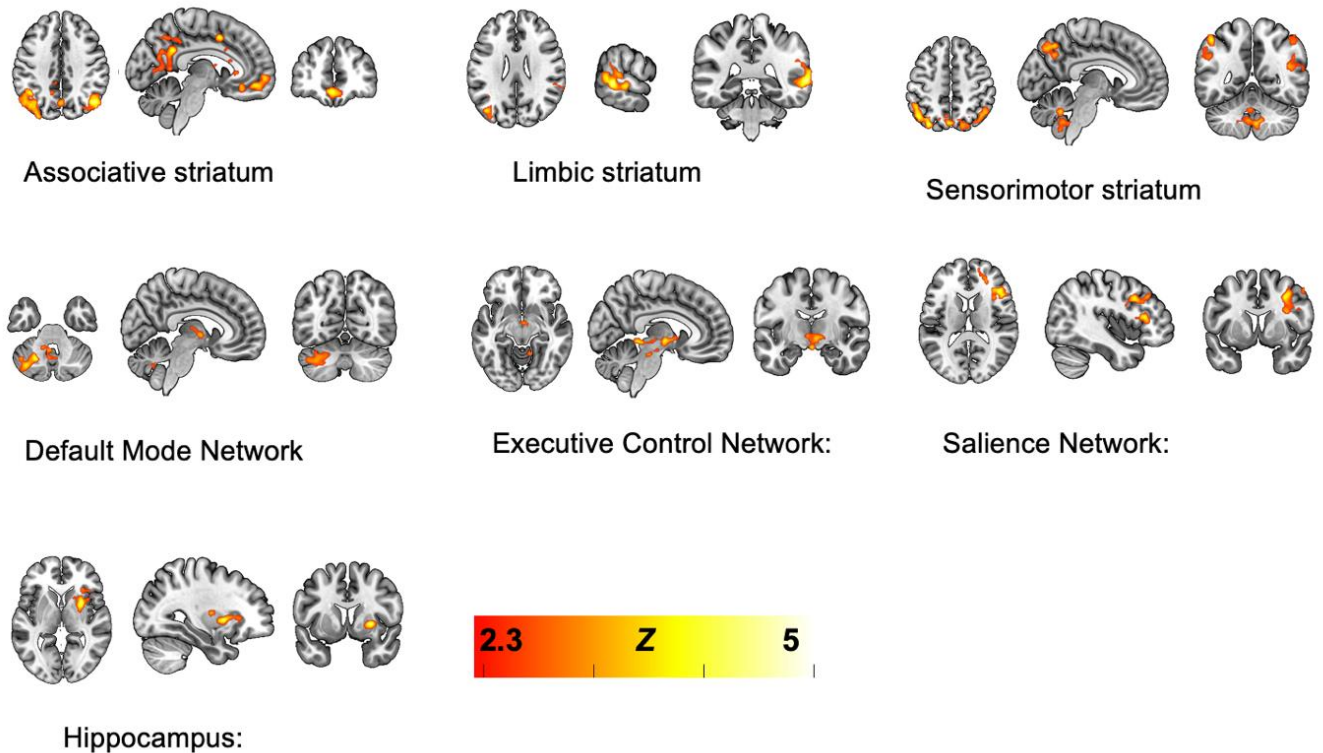


Figure 4: Areas identified in all networks which differed significantly depending on age group. Background image is the MNI152 standard template brain. Images in neurological format (left of the image = left hemisphere).

### Network ROIs

The group-mean (all subjects, all scans) results were used to produce whole-network masks, with data from these masks providing a mean connectivity measure across each network, for analysis using conventional statistical methods. Results from these analyses are shown in supplementary figures 2 and 3, with the derived ROI masks shown in supplementary figure 4. The overall pattern seen for all the networks in these group-mean analyses conforms to previous work (Wall, Pope, Freeman, Kowalczyk, Demetriou, Mokrysz, Hindocha, Lawn, Bloomfield, Freeman, Feilding, D. Nutt, *et al.*, 2019) (Wall *et al.*, 2020), so these analyses also serve to validate the acquisition and analysis methods used.

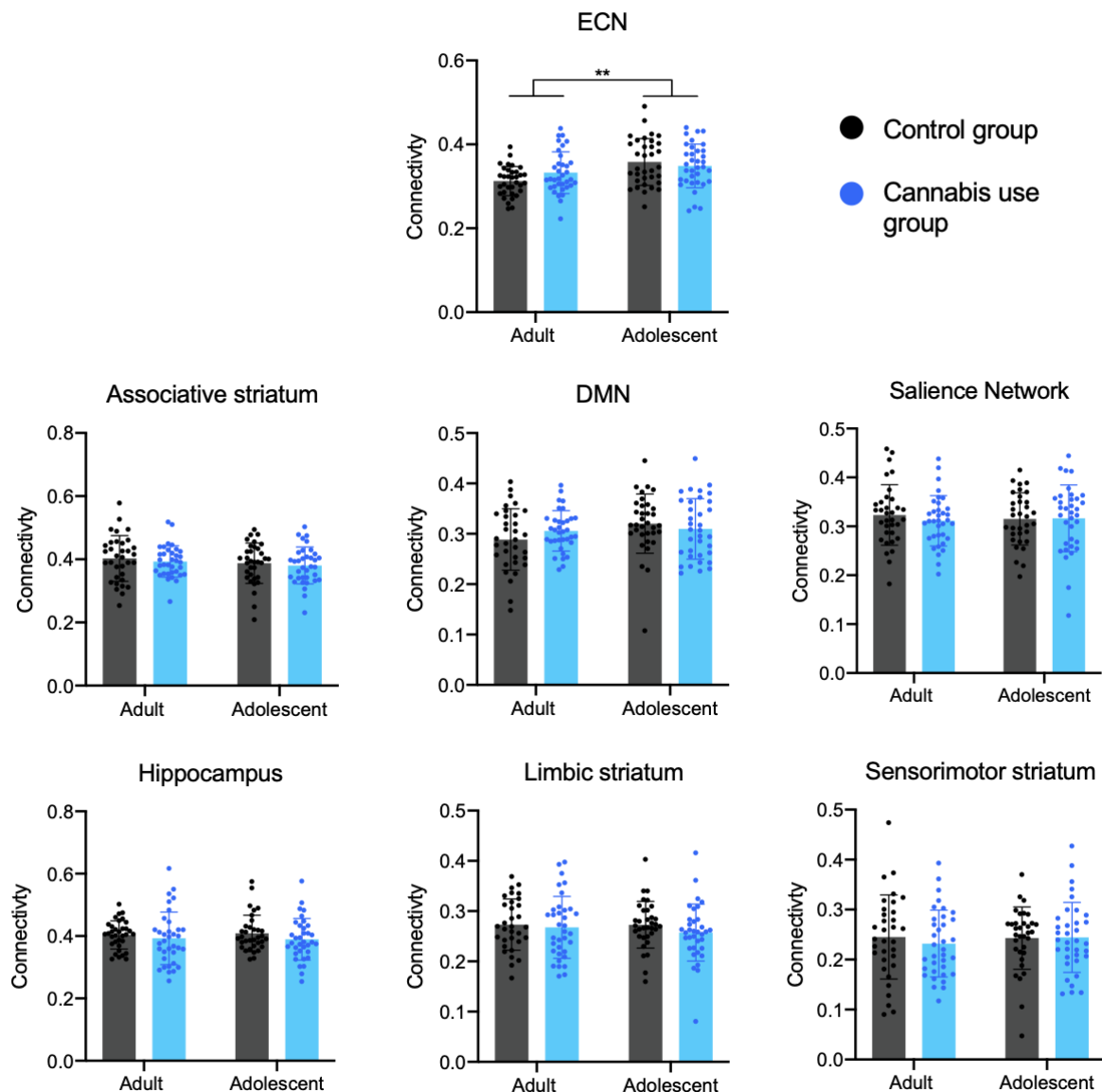


Figure 5: Network ROI analysis results. No effects of regular cannabis use or interaction effects were found. An age group effect (adolescents > adults) was identified in the Executive Control Network (ECN) only:  $F(1, 134) = 13.88, P=0.0003, n=138$ .

From the seven networks tested (Salience, DMN, ECN, associative striatum, limbic striatum, sensorimotor striatum, and hippocampus) none showed a difference between cannabis use-groups compared to control-groups; Bayesian *post hoc* tests had Bayes factors  $BF_{10} < 0.33$  indicating moderate evidence (Associative striatum, DMN, ECN, Salience, Sensorimotor striatum) and  $BF_{10} < 1$  anecdotal evidence (Limbic and Hippocampal) for the null ( $H_0$ ). Only the ECN showed an age group effect (Figure 4),  $F(1, 134) = 13.88, P=0.0003, n=138$

(adolescents > adults), with a  $BF_{10}$  of 73, indicating strong evidence for the experimental hypothesis ( $H_1$ ). All other networks had a  $BF_{10} < 0.50$  indicating moderate evidence for the null ( $H_0$ ). Outliers were removed using the ROUT method ( $Q=1\%$ ). This left  $N=136$  in the Limbic striatum analysis, and  $N=137$  in the Sensorimotor striatum, DMN and SAL networks. The remaining networks had  $N=138$  in these analyses. Removal of outliers did not change the overall significance of any network. There were no significant interaction effects found in any of the whole network analyses, and there was very strong evidence for the null ( $H_0$ ) in the interaction terms in the associative striatum, DMN, hippocampus, salience and sensorimotor striatum networks ( $BF_{10} < 0.03$ ) and some evidence for the null in the limbic striatum network ( $BF_{10} < 0.10$ ). There was some evidence for the interaction hypothesis ( $H_1$ ) in the ECN ( $BF_{10}=15$ ) but this was not conventionally statistically significant  $F(1,134)=3.117$ ,  $P=0.08$ .

#### *Interaction (age x user-group) effects*

There were no significant interaction effects between age and cannabis use-groups found in the seed-to-voxel analyses (no significant clusters in the group-level comparisons). There were also no significant interaction effects present in the analyses of the network ROI data.

#### *Correlation with cannabis use*

Correlation analyses were conducted to investigate the relationships between cannabis use frequency, and each network-ROI derived data, separately for each cannabis use-group. A Bonferroni correction for multiple tests applied to the 14 correlations conducted yielded a corrected alpha value of 0.004. There were no significant correlations present in these analyses at this threshold. These findings are summarised in table 5.

---

**Network**

**Frequency of cannabis use**

	<b>Adults</b>	<b>Adolescents</b>
Saliency Network	r = -0.158	r= -0.039
<i>Anterior Insula seed</i>	p=0.364	p= 0.826
Associative striatum	r= -0.217	r= 0.267
	p= 0.211	p= 0.121
Hippocampal Network	r= -0.156	r= -0.391
	p= 0.371	p= 0.02
Limbic Striatum	r= -0.227	r= -0.064
	p= 0.190	p= 0.717
Default Mode Network	r= 0.104	r= 0.178
<i>Posterior Cingulate seed</i>	p= 0.552	p= 0.306
Sensorimotor Striatum	r= -0.064	r= 0.074
	p= 0.717	p= 0.671
Executive Control Network	r= -0.025	r= -0.109
<i>Dorsolateral prefrontal cortex seed</i>	p= 0.888	p= 0.534

Table 5: Summary statistics of the correlations conducted (using Pearson correlation) to investigate the relationship between overall resting state network connectivity and frequency of cannabis use. None of the results are significant at a Bonferroni-corrected alpha threshold of  $p < 0.004$ .

Although not in our original preregistered analysis plan, we also thought it important to test for an association between life-time cannabis use and network connectivity since we identified a significant difference between lifetime cannabis uses in adolescents and adult cannabis use groups (table 4). We found no significant correlations between any of the networks and life-time cannabis use (table 6).

<b>Network</b>	<b>r value</b>	<b>P value</b>
Saliency network	-0.123	0.315
Associative striatal network	0.059	0.625
Hippocampal network	-0.138	0.250
Limbic striatal network	-0.076	0.530
Default mode network	-0.002	0.982
Sensorimotor striatal network	0.0364	0.766
Executive control network	0.0048	0.690

Table 6: Summary statistics of the correlations conducted (using Pearson correlation) to investigate the relationship between overall resting state network connectivity and lime-time cannabis uses. None of the results are significant at a Bonferroni-corrected alpha threshold of  $p < 0.004$ .

## Discussion

We have identified differences in resting-state functional connectivity of the Executive Control Network (ECN) associated with cannabis use in a large cohort of adolescent and adult cannabis users and age- and gender-matched controls. Seed-to-voxel analyses showed regional differences between the cannabis use group and the control group in ECN connectivity, but not in any other network tested (the DMN, salience network, hippocampal network, and three striatal networks). All areas identified in the ECN analysis showed an increase in functional connectivity in the cannabis use groups compared to the control groups. Examining overall network changes using network-mask ROIs showed no effects of cannabis use-group, but an effect of age group (adolescents > adults), also in the ECN. Crucially, there were no significant age-group by cannabis use-group interaction effects, suggesting the relationship between regular cannabis use and RSN functional connectivity is not different in adolescents and adults, contrary to our original hypothesis. Our study does not provide any evidence that adolescent cannabis users are more vulnerable to the putatively harmful impacts of regular cannabis use on the brain's functional connectivity. Also contrary to our hypotheses, there were no significant relationships found between cannabis use frequency and functional connectivity measures.

Effects of age-group were also evident in many of the networks (see figure 4 and the supplementary material for details). With only one exception, the regions identified show relatively decreased connectivity in adolescents; consistent with a developmental/maturational trajectory of increasing resting-state functional connectivity in cortical networks (van Duijvenvoorde *et al.*, 2016) and other work suggesting differentiations in the reward system in adolescents and adults (Telzer, 2016). As these results are largely confirmatory and were not the focus of this study, they will not be discussed further.

ECN connectivity changes associated with cannabis use have been previously documented. Relative increases in connectivity between the prefrontal and parietal cortices have been shown in adolescent abstinent cannabis users compared to controls (Blest-Hopley, Giampietro and Bhattacharyya, 2019). It has been suggested that the increases in connectivity in regions within the network which mediates control may be compensatory for the relative cognitive/connectivity impairments caused by regular cannabis intoxication (Harding *et al.*, 2012). Consistent with this interpretation, previous data has shown reductions in ECN connectivity in acute dosing experiments (Wall *et al.*, 2019). In the current data we saw no cannabis use-group effects on the whole network ROI analysis, but we did see relative increases in connectivity in localised regions close to, but largely not overlapping, the network. There was an increase in connectivity to the anterior cingulate and supplementary motor regions which are slightly posterior to the region of the cingulate in the canonical ECN (see figure 2). Moreover, we saw increases in connectivity in the TPJ region

and superior temporal gyrus; not regions normally associated with the ECN and executive functions. Therefore, these data may represent a possible allostatic or compensatory response to impairments produced by acute cannabis use. However, because there is also some migration of the network to regions outside the canonical network, an alternative interpretation is that this may be evidence of dysfunction. In particular, the increased connectivity with the ECN of the anterior cingulate cortex and insula are noteworthy, as these regions have been implicated in the development of compulsive drug use and addiction (ACC: Peoples, 2002; Zhao *et al.*, 2021 insula: Naqvi and Bechara, 2009; Droutman, Read and Bechara, 2015). Further work should look at these networks while participants complete a relevant executive task to investigate if the relative change in network positioning/connectivity affects task performance. In the analyses of whole-network ROI masks (figure 5) we saw significantly lower connectivity in the adult compared to the adolescent group, with a trend-level interaction effect ( $P = 0.08$ ). The overall picture is therefore: increased connectivity *within* the network in adolescents, but also increased connectivity to regions outside the network in the cannabis use groups.

We found no consistent differences associated with cannabis use on DMN connectivity. The DMN is comparatively well-studied, but some previous authors have reported decreases in connectivity associated with cannabis use (Wetherill *et al.*, 2015), while others have reported increases (Pujol *et al.*, 2014). At least some of the variability in results may be related to the seed-regions used to define the DMN. Previous work has shown that the DMN may be meaningfully fractionated with the use of relatively more dorsal/anterior or ventral/posterior PCC/precuneus seeds (Zhang and Li, 2012; Chen *et al.*, 2017). Connectivity of the ventral portion may be more tightly involved with internally-directed attention, while the dorsal division is more integrated with cognitive control networks, and therefore may play a role in modulating activity between networks to cope with external task demands (Leech *et al.*, 2011; Leech and Sharp, 2014). The seed-region used here (derived using meta-analytic data from <https://neurosynth.org/>) is localised to the ventral portion and is therefore more likely to represent connectivity associated with internally-directed attention and processes.

Previous work suggests cannabis use can cause the DMN to become less active during rest and more active while engaging in a task, with the reduced deactivation during an active task correlating with lower performance (Bossong *et al.*, 2013). Since the DMN and ECN tend to work in opposition (Fox *et al.*, 2005) the current results for ECN connectivity may also reflect this dysfunctional process. A related interpretation is that these results may represent a decrease in the brain's modularity. Modularity is the tendency of the brain to form well-defined networks with relatively strong connectivity within networks and relatively weak connectivity between networks. This might explain the relative increases in connectivity with



the non-DMN associated areas in previous work, and the non-ECN associated areas in the current results. This is an emerging concept in work on classic psychedelics, has been shown with acute challenge studies (Petri *et al.*, 2014), and may also be related to clinical anti-depressive effects of psychedelic therapy (Daws *et al.*, 2021). Further work will be needed with specialised analysis techniques to investigate this possibility.

Striatal connectivity has been less frequently examined in previous work, despite a high expression of CB<sub>1</sub> receptors in the striatum and other sub-cortical regions (Svíženská, Dubový and Šulcová, 2008). Positron Emission Tomography (PET) studies have shown that cannabinoids may act as a modulator of dopamine release in the striatum (Bossong *et al.*, 2009; Calakos *et al.*, 2021) and that cannabis users may have reduced dopamine synthesis capacity (Bloomfield *et al.*, 2014). However, other similar studies have shown no effect on dopamine release (Stokes *et al.*, 2009) and that life-time use of cannabis is not associated with dopamine receptor availability in the limbic striatum (Stokes *et al.*, 2012). Further understanding of the long-term effects of cannabis on the striatum (and by implication, dopaminergic systems) is essential because of the key role these areas play in cannabis-related addiction and psychosis (Curran *et al.*, 2016). In the present results, no differences related to cannabis use status were identified in the striatal networks investigated, or the hippocampus. This is encouraging, as it suggests that (at least in the present sample) cannabis use is not significantly affecting striato-cortical connectivity.

These results indicate that the cannabis use groups have somewhat different patterns of functional connectivity to the control groups, however there were no significant interactions between cannabis use and age-group, or in other words, the differences between the cannabis use groups and the control groups were similar in adults and adolescents. This is contrary to our original hypotheses and rationale for the study; that cannabis may affect cortico-striatal development and may present greater disruption of networks compared to non-user controls in the adolescent group. In addition, we also found no significant correlations between cannabis frequency of use and any derived connectivity measure. Recent systematic reviews focussed on the question of cannabis use in adolescence have been somewhat inconclusive, based on the weak-to-moderate available evidence (Gorey *et al.*, 2019) and the small number of available studies (Blest-Hopley, Giampietro and Bhattacharyya, 2018). Pre-clinical work (Quinn *et al.*, 2008) showing later deficits associated with adolescent exposure is clearly still a cause for concern, however it appears that in humans the effects of adolescent and adult usage are not as clear-cut (at least, on the measures used here).

This study has a number of strengths. It is the largest investigation into cannabis related functional connectivity using neuroimaging to examine a cannabis use group and a matched-

control group to date. Furthermore, the design is highly novel; no existing fMRI studies have directly compared adolescent and adult cannabis users with age-matched controls. Additionally, our controls were also matched on sex and the adult/adolescent cannabis users were matched on cannabis use frequency, at ~4 days/week. Drug and alcohol abstinence were also biologically verified.

Interaction effects are generally smaller and require more statistical power to substantiate (Marshall, 2007). It is possible that the sample was still under-powered to detect a hypothetical interaction, of unknown effect size, between age and cannabis use-groups. The conservative correction for multiple comparisons used in the seed-to-voxel analyses may have reduced our ability to detect an interaction, if it did exist. However, mitigating against this are the results from the overall network-mask analyses, which also showed no interaction effects. These overall measures of connectivity within a network are somewhat crude in that they are a single summary measure averaged over an entire network, however they are sensitive in that the conservative correction for multiple comparisons required for seed-to-voxel analyses is not necessary. Another important limitation is related to the observational and cross-sectional nature of the study design. We cannot establish causality, and ideally, additional longitudinal analyses are needed to rule out alternative explanations for the effects seen (e.g. possibly confounding effects of genetics, environmental or social factors, or any other pre-existing difference in the groups; Hicks *et al.*, 2013). Despite our best efforts to recruit carefully matched groups, there were group differences on some possible confounding measures such as use of other drugs (with the cannabis user group reporting significantly more than the control group) and risk-taking (adolescents > adults). The adult cannabis use group also had a significantly greater life-time cannabis use than the adolescent cannabis use group. Given this could have impacted our results we conducted further exploratory correlations with life-time cannabis use and network connectivity. No correlations were found, which suggests this was not a confound in our results, although it still cannot be ruled out entirely. The borderline-significant ( $p = 0.046$ ) difference in educational level between adult users and controls (controls > users) is also worth noting given the effects seen in the ECN, a network known to be involved in working memory and other complex cognitive functions (Rottschy *et al.*, 2012). The influence of these potential confounds is unclear, however additional longitudinal data may also be helpful in this regard. Moreover, in both the adolescent and adult cannabis use groups, the age of first weekly use often started before the age of 25 when brain development is incomplete (Arain *et al.*, 2013). Therefore, differences in functional connectivity between adolescents and adults may be due to effects of cannabis on brain development rather than effects of cannabis use *per se*. This is an inherent problem of human research and was almost impossible to control for while trying to obtain a sufficiently large sample size.

Our cannabis-use group was heterogeneous in terms of frequency of usage at 1-7 days/week, however significant correlations with use frequency were not found. The influence of sex on the effects of cannabis use is documented (e.g. McPherson *et al.*, 2021), and may be an important factor in both acute and long-term response to cannabis. This is particularly relevant when researching adolescents since girls tend to undergo earlier maturation than boys (Lenroot *et al.*, 2007); this suggests male brains may be most vulnerable to changes around the time that cannabis is often first used in the adolescent population. Animal studies have also suggested that the density of CB1 receptors may be different between males and females. Earlier in life males are thought to have a greater density of CB1 receptors than females (Mateos *et al.*, 2011) however this is thought to change towards the later years of life (Van Laere *et al.*, 2008). The interactions with hormones are also important to consider as there is evidence for varying levels of CB1 receptors across the oestrus cycle (Bradshaw *et al.*, 2006). Though there is animal evidence suggesting a potential difference between sexes, there is little human evidence so far. One study identified a difference in the reported usage and side effects of cannabis use between men and women, for example, men were more likely to consume more and report increased musicality, while women were more likely to have a desire to clean (Cuttler, Mischley and Sexton, 2016). While the animal evidence indicates there may be some sex differences the human research is not quite as compelling, furthermore, there may be some underlying gender-stereotyped and misogynistic undertones to this work; the ethical implications of studying these difference needs to be carefully addressed. Our sample was balanced as closely as possible within each group for biological sex, but sub-dividing our groups further to explicitly examine differences between males and females (and potential interactions of sex, age, and user groups) would mean under-powered and thus unreliable analyses.

To summarise, we found a significant increase in ECN connectivity in the cannabis use group compared to the control group. This pattern of regional increases in functional connectivity in the non-intoxicated state may reflect adaptive allostatic or compensatory processes, arising in response to the regular acute disruption of the ECN by regular cannabis use. However, given that the areas of increased connectivity were not closely overlapping the canonical ECN, these data may alternatively be evidence for dysfunction. No interactions between age and cannabis-use groups were found, suggesting that these effects are similar in both age-groups, and adolescents are not hyper-vulnerable to cannabis-related alterations to resting state networks. No correlations were present between network function and cannabis use frequency. Although most networks were unaffected, regular cannabis use appears to have some long-term effects on resting-state brain function and future work should focus on substantiating these results further with longitudinal studies

and investigating the implications of these changes for general cognitive and emotional function, as well as the development of pathological states.

## **Data availability**

The datasets generated during and/or analysed during the current study are available from the corresponding author on reasonable request

## **Author contributions**

Conception and methodology: WL, CM, TF, EV, HVC, MBW

Funding acquisition: HVC, TF

Data collection: WL, CM, NA, AB, NFV, RL, SO, KP, KT

Data analysis: NE, MBW, KT

Supervision: HVC, MBW

Manuscript first draft: NE, MBW

All authors edited and approved the final manuscript.

## **Funding**

This study was funded by a grant from the Medical Research Council, MR/P012728/1, to HVC and TPF.

## **Competing interests**

HVC has consulted for Janssen Research and Development. MBW and NE's primary employer is Invicro LLC, a contract research organisation that performs commercial research for the pharmaceutical and biotechnology industries. The remaining authors have no conflicts of interest to disclose.

## **References**

Albaugh, M.D. *et al.* (2021) 'Association of Cannabis Use During Adolescence With Neurodevelopment', *JAMA Psychiatry*, 78(9), pp. 1031–1040. Available at: <https://doi.org/10.1001/JAMAPSYCHIATRY.2021.1258>.

- Arain, M. *et al.* (2013) ‘Maturation of the adolescent brain’, *Neuropsychiatric Disease and Treatment*, 9, pp. 449–461. Available at: <https://doi.org/10.2147/NDT.S39776>.
- Behzadi, Y. *et al.* (2007) ‘A component based noise correction method (CompCor) for BOLD and perfusion based fMRI.’, *NeuroImage*, 37(1), pp. 90–101. Available at: <https://doi.org/10.1016/j.neuroimage.2007.04.042>.
- Blakemore, S.J. (2008) ‘The social brain in adolescence’, *Nature Reviews Neuroscience* 2008 9:4, 9(4), pp. 267–277. Available at: <https://doi.org/10.1038/nrn2353>.
- Blest-Hopley, G., Giampietro, V. and Bhattacharyya, S. (2018) ‘Residual effects of cannabis use in adolescent and adult brains — A meta-analysis of fMRI studies’, *Neuroscience and Biobehavioral Reviews*, 88, pp. 26–41. Available at: <https://doi.org/10.1016/j.neubiorev.2018.03.008>.
- Blest-Hopley, G., Giampietro, V. and Bhattacharyya, S. (2019) ‘Regular cannabis use is associated with altered activation of central executive and default mode networks even after prolonged abstinence in adolescent users: Results from a complementary meta-analysis’, *Neuroscience and Biobehavioral Reviews*, 96, pp. 45–55. Available at: <https://doi.org/10.1016/j.neubiorev.2018.10.026>.
- Bloomfield, M.A.P. *et al.* (2014) ‘Dopaminergic function in cannabis users and its relationship to cannabis-induced psychotic symptoms’, *Biological Psychiatry*, 75(6), pp. 470–478. Available at: <https://doi.org/10.1016/j.biopsych.2013.05.027>.
- Bossong, M.G. *et al.* (2009) ‘ $\Delta$ 9-Tetrahydrocannabinol Induces Dopamine Release in the Human Striatum’, *Neuropsychopharmacology*, 34(3), pp. 759–766. Available at: <https://doi.org/10.1038/npp.2008.138>.
- Bossong, M.G. *et al.* (2013) ‘Default Mode Network in the Effects of  $\Delta$ 9-Tetrahydrocannabinol (THC) on Human Executive Function’, *PLoS ONE*, 8(7). Available at: <https://doi.org/10.1371/JOURNAL.PONE.0070074>.
- Bradshaw, H.B. *et al.* (2006) ‘Sex and hormonal cycle differences in rat brain levels of pain-related cannabimimetic lipid mediators’, *American Journal of Physiology. Regulatory, Integrative and Comparative Physiology*, 291(2), pp. R349-358. Available at: <https://doi.org/10.1152/ajpregu.00933.2005>.
- Calakos, K.C. *et al.* (2021) ‘Assessment of transient dopamine responses to smoked cannabis’, *Drug and Alcohol Dependence*, 227, p. 108920. Available at: <https://doi.org/10.1016/J.DRUGALCDEP.2021.108920>.
- Chan G. C. K., Hinds T. R., Impey, S., & Storm, D.R. (1998) ‘Hippocampal neurotoxicity of  $\Delta$ 9-tetrahydrocannabinol.’, *Journal of Neuroscience*, 18, pp. 5322–5332.
- Chen, J.E. *et al.* (2017) ‘Dissociated patterns of anti-correlations with dorsal and ventral default-mode networks at rest’, *Human Brain Mapping*, 38(5), pp. 2454–2465. Available at: <https://doi.org/10.1002/hbm.23532>.
- Comninos, A.N. *et al.* (2018) ‘Modulations of human resting brain connectivity by kisspeptin enhance sexual and emotional functions’, *JCI insight*, 3(20), pp. 0–10. Available at: <https://doi.org/10.1172/jci.insight.121958>.

- Curran, H.V. *et al.* (2016) ‘Keep off the grass? Cannabis, cognition and addiction’, *Nature Reviews Neuroscience*, 17(5), pp. 293–306. Available at: <https://doi.org/10.1038/nrn.2016.28>.
- Cuttler, C., Mischley, L.K. and Sexton, M. (2016) ‘Sex Differences in Cannabis Use and Effects: A Cross-Sectional Survey of Cannabis Users’, *Cannabis and Cannabinoid Research*, 1(1), pp. 166–175. Available at: <https://doi.org/10.1089/can.2016.0010>.
- Daws, R. *et al.* (2021) ‘Decreased brain modularity after psilocybin therapy for depression’, *PREPRINT (Version 1) available at Research Square [https://doi.org/10.21203/rs.3.rs-513323/v1]*, pp. 1–19.
- Degenhardt, L. *et al.* (2016) ‘The increasing global health priority of substance use in young people’, *The Lancet Psychiatry*, 3(3), pp. 251–264. Available at: [https://doi.org/10.1016/S2215-0366\(15\)00508-8](https://doi.org/10.1016/S2215-0366(15)00508-8).
- Demetriou, L. *et al.* (2018) ‘A comprehensive evaluation of increasing temporal resolution with multiband-accelerated sequences and their effects on statistical outcome measures in fMRI’, *NeuroImage*, 176, pp. 404–416.
- Desikan, R.S. *et al.* (2006) ‘An automated labeling system for subdividing the human cerebral cortex on MRI scans into gyral based regions of interest’, *NeuroImage*, 31(3), pp. 968–980. Available at: <https://doi.org/10.1016/j.neuroimage.2006.01.021>.
- Downer, E. *et al.* (2001) ‘ $\Delta 9$ -Tetrahydrocannabinol induces the apoptotic pathway in cultured cortical neurones via activation of the CB1 receptor’, *NeuroReport*, 12(18), pp. 3973–3978. Available at: <https://doi.org/10.1097/00001756-200112210-00024>.
- Droutman, V., Read, S.J. and Bechara, A. (2015) ‘Revisiting the role of the insula in addiction’, *Trends in Cognitive Sciences*, 19(7), pp. 414–420. Available at: <https://doi.org/10.1016/j.tics.2015.05.005>.
- van Duijvenvoorde, A.C.K. *et al.* (2016) ‘Testing a dual-systems model of adolescent brain development using resting-state connectivity analyses’, *NeuroImage*, 124, pp. 409–420. Available at: <https://doi.org/10.1016/j.neuroimage.2015.04.069>.
- Fox, M.D. *et al.* (2005) ‘The human brain is intrinsically organized into dynamic, anticorrelated functional networks.’, *Proceedings of the National Academy of Sciences of the United States of America*, 102(27), pp. 9673–8. Available at: <https://doi.org/10.1073/pnas.0504136102>.
- Fox, M.D. and Raichle, M.E. (2007) ‘Spontaneous fluctuations in brain activity observed with functional magnetic resonance imaging.’, *Nature reviews. Neuroscience*, 8(9), pp. 700–11. Available at: <https://doi.org/10.1038/nrn2201>.
- Friston, K.J. *et al.* (1994) ‘Assessing the significance of focal activations using their spatial extent’, *Human Brain Mapping*, 1(3), pp. 210–220. Available at: <https://doi.org/10.1002/hbm.460010306>.
- Gorey, C. *et al.* (2019) ‘Age-related differences in the impact of cannabis use on the brain and cognition: a systematic review’, *European Archives of Psychiatry and Clinical Neuroscience*, 269, pp. 37–58. Available at: <https://doi.org/10.1007/s00406-019-00981-7>.

- Goulden, N. *et al.* (2014) ‘The salience network is responsible for switching between the default mode network and the central executive network: Replication from DCM’, *NeuroImage*, 99, pp. 180–190. Available at: <https://doi.org/10.1016/j.neuroimage.2014.05.052>.
- Harding, I.H. *et al.* (2012) ‘Functional Connectivity in Brain Networks Underlying Cognitive Control in Chronic Cannabis Users’, *Neuropsychopharmacology*, 37(8), pp. 1923–1933. Available at: <https://doi.org/10.1038/npp.2012.39>.
- Heath, R.G. *et al.* (1980) ‘Cannabis sativa: Effects on brain function and ultrastructure in rhesus monkeys’, *Biological Psychiatry*, 15(5), pp. 657–690.
- Hicks, B.M. *et al.* (2013) ‘Gene-Environment Correlation in the Development of Adolescent Substance Abuse: Selection Effects of Child Personality and Mediation via Contextual Risk Factors’, *Development and psychopathology*, 25(1), p. 119. Available at: <https://doi.org/10.1017/S0954579412000946>.
- Jacobus, J. and Tapert, S. (2014) ‘Effects of Cannabis on the Adolescent Brain’, *Current Pharmaceutical Design*, 20(13), pp. 2186–2193. Available at: <https://doi.org/10.2174/13816128113199990426>.
- Joel, D. and Weiner, I. (2000) ‘The Connections of the Dopaminergic System With the Striatum in Rats and Primates : an Analysis With Respect To the Functional and Compartmental Organization of the Striatum’, *Neuroscience*, 96(3), pp. 451–474.
- Lawn, W. *et al.* (2019) ‘cannTEEN: How does long-term cannabis use affect teenagers’ and adults’ cognition, mental health and brains?’, p. 17. Available at: <https://doi.org/none>.
- Lawn, Will *et al.* (2022) ‘The CannTeen Study: Cannabis use disorder, depression, anxiety, and psychotic-like symptoms in adolescent and adult cannabis users and age-matched controls’, *Journal of Psychopharmacology*, pp. 1–12. Available at: <https://doi.org/10.31234/OSF.IO/C7X95>.
- Lawn, W *et al.* (2022) ‘The CannTeen study: verbal episodic memory, spatial working memory, and response inhibition in adolescent and adult cannabis users and age-matched controls’, *Psychopharmacology 2022*, 1, pp. 1–13. Available at: <https://doi.org/10.1007/S00213-022-06143-3>.
- Leech, R. *et al.* (2011) ‘Fractionating the Default Mode Network: Distinct Contributions of the Ventral and Dorsal Posterior Cingulate Cortex to Cognitive Control’, *Journal of Neuroscience*, 31(9), pp. 3217–3224. Available at: <https://doi.org/10.1523/JNEUROSCI.5626-10.2011>.
- Leech, R. and Sharp, D.J. (2014) ‘The role of the posterior cingulate cortex in cognition and disease’, *Brain*, 137(1), pp. 12–32. Available at: <https://doi.org/10.1093/brain/awt162>.
- Lenroot, R.K. *et al.* (2007) ‘Sexual dimorphism of brain developmental trajectories during childhood and adolescence’, *NeuroImage*, 36(4), pp. 1065–1073. Available at: <https://doi.org/10.1016/j.neuroimage.2007.03.053>.
- Lichenstein, S.D. *et al.* (2022) ‘Systematic review of structural and functional neuroimaging studies of cannabis use in adolescence and emerging adulthood: evidence from 90 studies and



9441 participants', *Neuropsychopharmacology*, 47(5), pp. 1000–1028. Available at: <https://doi.org/10.1038/s41386-021-01226-9>.

Marshall, S.W. (2007) 'Power for tests of interaction: Effect of raising the Type I error rate', *Epidemiologic Perspectives and Innovations*, 4(1), pp. 1–7. Available at: <https://doi.org/10.1186/1742-5573-4-4/TABLES/1>.

Martinez, D. *et al.* (2003) 'Imaging human mesolimbic dopamine transmission with positron emission tomography. Part II: amphetamine-induced dopamine release in the functional subdivisions of the striatum.', *Journal of cerebral blood flow and metabolism : official journal of the International Society of Cerebral Blood Flow and Metabolism*, 23(3), pp. 285–300. Available at: <https://doi.org/10.1097/01.WCB.0000048520.34839.1A>.

Mateos, B. *et al.* (2011) 'Adolescent exposure to nicotine and/or the cannabinoid agonist CP 55,940 induces gender-dependent long-lasting memory impairments and changes in brain nicotinic and CB(1) cannabinoid receptors', *Journal of Psychopharmacology (Oxford, England)*, 25(12), pp. 1676–1690. Available at: <https://doi.org/10.1177/0269881110370503>.

McPherson, K.L. *et al.* (2021) 'Cannabis Affects Cerebellar Volume and Sleep Differently in Men and Women', *Frontiers in Psychiatry*, 12. Available at: <https://www.frontiersin.org/articles/10.3389/fpsy.2021.643193> (Accessed: 25 October 2022).

Meyer, H.C., Lee, F.S. and Gee, D.G. (2017) 'The Role of the Endocannabinoid System and Genetic Variation in Adolescent Brain Development', *Neuropsychopharmacology* 2018 43:1, 43(1), pp. 21–33. Available at: <https://doi.org/10.1038/npp.2017.143>.

Naqvi, N.H. and Bechara, A. (2009) 'The hidden island of addiction: the insula', *Trends in Neurosciences*, 32(1), pp. 56–67. Available at: <https://doi.org/10.1016/j.tins.2008.09.009>.

Nestor, L., Hester, R. and Garavan, H. (2010) 'Increased ventral striatal BOLD activity during non-drug reward anticipation in cannabis users', *NeuroImage*, 49(1), pp. 1133–1143. Available at: <https://doi.org/10.1016/j.neuroimage.2009.07.022>.

Office for National Statistics (2022) *Drug misuse in England and Wales -*. Available at: <https://www.ons.gov.uk/peoplepopulationandcommunity/crimeandjustice/articles/drugmisuseinenglandandwales/yearendingjune2022> (Accessed: 16 January 2023).

Paronis, C.A. *et al.* (2012) 'Δ9-Tetrahydrocannabinol acts as a partial agonist/antagonist in mice', *Behavioural pharmacology*, 23(8), p. 802. Available at: <https://doi.org/10.1097/FBP.0B013E32835A7C4D>.

Peoples, L.L. (2002) 'Will, Anterior Cingulate Cortex, and Addiction', *Science*, 296(5573), pp. 1623–1624. Available at: <https://doi.org/10.1126/science.1072997>.

Petri, G. *et al.* (2014) 'Homological scaffolds of brain functional networks', *Journal of The Royal Society Interface*, 11(101). Available at: <https://doi.org/10.1098/RSIF.2014.0873>.

Pujol, J. *et al.* (2014) 'Functional connectivity alterations in brain networks relevant to self-awareness in chronic cannabis users', *Journal of Psychiatric Research*, 51(1), pp. 68–78. Available at: <https://doi.org/10.1016/j.jpsychires.2013.12.008>.

Quinn, H.R. *et al.* (2008) 'Adolescent rats find repeated Δ9-THC less aversive than adult rats but display greater residual cognitive deficits and changes in hippocampal protein expression

following exposure’, *Neuropsychopharmacology*, 33(5), pp. 1113–1126. Available at: <https://doi.org/10.1038/sj.npp.1301475>.

Raichle, M.E. (2015) ‘The Brain’s Default Mode Network’. Available at: <https://doi.org/10.1146/annurev-neuro-071013-014030>.

Ritchay, M.M. *et al.* (2021) ‘Resting state functional connectivity in the default mode network: Relationships between cannabis use, gender, and cognition in adolescents and young adults’, *NeuroImage: Clinical*, 30, p. 102664. Available at: <https://doi.org/10.1016/J.NICL.2021.102664>.

Rodríguez de Fonseca, F. *et al.* (1993) ‘Presence of cannabinoid binding sites in the brain from early postnatal ages’, *Neuroreport*, 4(2), pp. 135–138. Available at: <https://doi.org/10.1097/00001756-199302000-00005>.

Rottschy, C. *et al.* (2012) ‘Modelling neural correlates of working memory: A coordinate-based meta-analysis’, *NeuroImage*, 60(1), pp. 830–846. Available at: <https://doi.org/10.1016/j.neuroimage.2011.11.050>.

Schnakenberg Martin, A.M. *et al.* (2021) ‘Altered cerebellar-cortical resting-state functional connectivity in cannabis users’, *Journal of Psychopharmacology (Oxford, England)*, 35(7), pp. 823–832. Available at: <https://doi.org/10.1177/02698811211019291>.

Seeley, W.W. *et al.* (2007) ‘Dissociable intrinsic connectivity networks for salience processing and executive control.’, *The Journal of neuroscience : the official journal of the Society for Neuroscience*, 27(9), pp. 2349–56. Available at: <https://doi.org/10.1523/JNEUROSCI.5587-06.2007>.

Skumlien, M. *et al.* (2021) ‘The acute and non-acute effects of cannabis on reward processing: A systematic review’, *Neuroscience & Biobehavioral Reviews*, 130, pp. 512–528. Available at: <https://doi.org/10.1016/J.NEUBIOREV.2021.09.008>.

Skumlien, M. *et al.* (2022) ‘ARTICLE Neural responses to reward anticipation and feedback in adult and adolescent cannabis users and controls’. Available at: <https://doi.org/10.1038/s41386-022-01316-2>.

Smith, S.M. *et al.* (2004) ‘Advances in functional and structural MR image analysis and implementation as FSL’, *Neuroimage*, 23, pp. S208–S219.

Solé-Padullés, C. *et al.* (2016) ‘Intrinsic connectivity networks from childhood to late adolescence: Effects of age and sex’, *Developmental Cognitive Neuroscience*, 17, pp. 35–44. Available at: <https://doi.org/10.1016/j.dcn.2015.11.004>.

Stevens, M.C., Pearlson, G.D. and Calhoun, V.D. (2009) ‘Changes in the interaction of resting-state neural networks from adolescence to adulthood’, *Human brain mapping*, 30(8), pp. 2356–2366. Available at: <https://doi.org/10.1002/HBM.20673>.

Stokes, P. *et al.* (2012) ‘History of cannabis use is not associated with alterations in striatal dopamine D2/D3 receptor availability’, *Journal of psychopharmacology (Oxford, England)*, 26(1), pp. 144–149. Available at: <https://doi.org/10.1177/02698811111414090>.

- Stokes, P.R.A. *et al.* (2009) ‘Can recreational doses of THC produce significant dopamine release in the human striatum?’, *NeuroImage*, 48(1), pp. 186–190. Available at: <https://doi.org/10.1016/J.NEUROIMAGE.2009.06.029>.
- Svíženská, I., Dubový, P. and Šulcová, A. (2008) ‘Cannabinoid receptors 1 and 2 (CB1 and CB2), their distribution, ligands and functional involvement in nervous system structures - A short review’, *Pharmacology Biochemistry and Behavior*, 90(4), pp. 501–511. Available at: <https://doi.org/10.1016/j.pbb.2008.05.010>.
- Teicher, M.H., Andersen, S.L. and Hostetter, J.C. (1995) ‘Evidence for dopamine receptor pruning between adolescence and adulthood in striatum but not nucleus accumbens’, *Developmental Brain Research*, 89(2), pp. 167–172. Available at: [https://doi.org/10.1016/0165-3806\(95\)00109-Q](https://doi.org/10.1016/0165-3806(95)00109-Q).
- Telzer, E.H. (2016) ‘Dopaminergic reward sensitivity can promote adolescent health: A new perspective on the mechanism of ventral striatum activation’, *Developmental Cognitive Neuroscience*, 17, pp. 57–67. Available at: <https://doi.org/10.1016/j.dcn.2015.10.010>.
- Thijssen, S. *et al.* (2017) ‘Regular cannabis and alcohol use is associated with resting-state time course power spectra in incarcerated adolescents’, *Drug and Alcohol Dependence*, 178, pp. 492–500. Available at: <https://doi.org/10.1016/j.drugalcdep.2017.05.045>.
- Thomas Yeo, B.T. *et al.* (2011) ‘The organization of the human cerebral cortex estimated by intrinsic functional connectivity’, *Journal of Neurophysiology*, 106(3), pp. 1125–1165. Available at: <https://doi.org/10.1152/jn.00338.2011>.
- Thomson, H. *et al.* (2022) ‘Is resting-state functional connectivity altered in regular cannabis users? A systematic review of the literature’, *Psychopharmacology*, 239(5), pp. 1191–1209. Available at: <https://doi.org/10.1007/s00213-021-05938-0>.
- Thurston, L. *et al.* (2022) ‘Melanocortin 4 receptor agonism enhances sexual brain processing in women with hypoactive sexual desire disorder’, *The Journal of Clinical Investigation*, 132(19). Available at: <https://doi.org/10.1172/JCI152341>.
- Tziortzi, A.C. *et al.* (2013) ‘Connectivity-Based Functional Analysis of Dopamine Release in the Striatum Using Diffusion-Weighted MRI and Positron Emission Tomography’. Available at: <https://doi.org/10.1093/cercor/bhs397>.
- Van Hell, H.H. *et al.* (2012) ‘Involvement of the endocannabinoid system in reward processing in the human brain’, *Psychopharmacology*, 219(4), pp. 981–990. Available at: <https://doi.org/10.1007/S00213-011-2428-8/TABLES/4>.
- Van Laere, K. *et al.* (2008) ‘Gender-dependent increases with healthy aging of the human cerebral cannabinoid-type 1 receptor binding using [(18)F]MK-9470 PET’, *NeuroImage*, 39(4), pp. 1533–1541. Available at: <https://doi.org/10.1016/j.neuroimage.2007.10.053>.
- Wall, M. *et al.* (2021) ‘Function of major cortical and striatal resting-state networks in adult and adolescent cannabis users’, p. 17. Available at: <https://doi.org/none>.
- Wall, M.B., Pope, R., Freeman, T.P., Kowalczyk, O.S., Demetriou, L., Mokrysz, C., Hindocha, C., Lawn, W., Bloomfield, M.A.P., Freeman, A.M., Feilding, A., Nutt, D.J., *et al.* (2019) ‘Dissociable effects of cannabis with and without cannabidiol on the human brain’s

resting-state functional connectivity’, *Journal of Psychopharmacology*, 33(7), pp. 822–830. Available at: <https://doi.org/10.1177/0269881119841568>.

Wall, M.B., Pope, R., Freeman, T.P., Kowalczyk, O.S., Demetriou, L., Mokrysz, C., Hindocha, C., Lawn, W., Bloomfield, M.A.P., Freeman, A.M., Feilding, A., Nutt, D., *et al.* (2019) ‘Dissociable effects of cannabis with and without cannabidiol on the human brain’s resting-state functional connectivity’, *Journal of Psychopharmacology*, 33(7), pp. 822–830. Available at: <https://doi.org/10.1177/0269881119841568>.

Wall, M.B. *et al.* (2020) ‘Individual and combined effects of Cannabidiol (CBD) and  $\Delta$ 9-tetrahydrocannabinol (THC) on striato-cortical connectivity in the human brain’, *bioRxiv*, p. 2020.11.20.391805. Available at: <https://doi.org/10.1101/2020.11.20.391805>.

Wall, M.B., Lam, C., *et al.* (2022) ‘Increased low-frequency brain responses to music after psilocybin therapy for depression’, *bioRxiv*, pp. 1–28.

Wall, M.B., Freeman, T.P., *et al.* (2022) ‘Individual and combined effects of Cannabidiol (CBD) and  $\Delta$ 9-tetrahydrocannabinol (THC) on striato-cortical connectivity in the human brain’, *Journal of Psychopharmacology*, p. 2020.11.20.391805. Available at: <https://doi.org/10.1177/02698811221092506>.

Wetherill, R.R. *et al.* (2015) ‘Cannabis, cigarettes, and their co-occurring use: Disentangling differences in default mode network functional connectivity’, *Drug and Alcohol Dependence*, 153, pp. 116–123. Available at: <https://doi.org/10.1016/j.drugalcdep.2015.05.046>.

Woo, C.W., Krishnan, A. and Wager, T.D. (2014) ‘Cluster-extent based thresholding in fMRI analyses: Pitfalls and recommendations’, *NeuroImage*, 91, pp. 412–419. Available at: <https://doi.org/10.1016/j.neuroimage.2013.12.058>.

Woolrich, M.W. *et al.* (2001) ‘Temporal autocorrelation in univariate linear modeling of FMRI data’, *NeuroImage*, 14(6), pp. 1370–1386. Available at: <https://doi.org/10.1006/nimg.2001.0931>.

Zhang, S. and Li, C.R. (2012) ‘Functional connectivity mapping of the human precuneus by resting state fMRI’, *NeuroImage*, 59(4), pp. 3548–3562. Available at: <https://doi.org/10.1016/j.neuroimage.2011.11.023>.

Zhao, Y. *et al.* (2021) ‘Anterior Cingulate Cortex in Addiction: New Insights for Neuromodulation’, *Neuromodulation: Technology at the Neural Interface*, 24(2), pp. 187–196. Available at: <https://doi.org/10.1111/ner.13291>.