# An opium curse? The long-run economic consequences of narcotics cultivation in British India

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#### **Abstract**

The long run consequences of colonial rule depended on the institutions introduced by the colonisers and the economic activities they promoted. This paper analyses the effects of opium production under British rule on current economic development in India. I employ a border discontinuity design which interacts fine-grained local variation in environmental suitability for poppy cultivation with administrative boundaries that demarcated opium-growing areas. I find that greater suitability for opium is associated with lower literacy and a lower rate of public good provision within opium-growing districts but has no effect in bordering areas where opium cultivation was prohibited. Placebo tests using suitability for other crops show no such discontinuity. Colonial administrative data allows me to test potential mechanisms for the persistent negative effect of opium production. Greater opium cultivation is associated with less per capita public spending on health and education by the British administration, a lower number of schools, and a greater concentration of police officers. These results suggest that colonial officials in opium growing districts concentrated on administering and policing the extraction of monopsony rents, while investing less in the wider local economy.

Keywords: colonialism, long-run development, resource curse

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# 1. Introduction

The state-sponsored export of opium from British India to China was arguably the largest and most enduring drug operation in history. At its peak in the mid-19<sup>th</sup> century it accounted for roughly 15% of total colonial revenue in India and 31% of India's exports (Richards 2002). To supply this trade the East India Company (EIC) – and later the British Government – developed a highly regulated cultivation system in which over one million farmers a year were under contract to grow opium poppies.<sup>2</sup> This paper evaluates the impact of this production on the long-term development of opium-growing areas and finds that it led to lower literacy and a lower rate of public good provision as measured by modern census data.

An influential literature in economics relates contemporary outcomes to the policies and institutions introduced by former colonial powers. Engerman and Sokoloff (2000), and Acemoglu, Johnson and Robinson (2001, 2002) argue that colonial rulers implemented different institutions depending on the factor endowments they encountered. Recent work has sought to build on these theories and add specificity by exploiting within-country variation in colonial policies and rich historical datasets not available at the cross-country level (Dell 2010, Dell and Olken 2017, Lowes and Montero 2017). While the British opium agencies were unique institutions – implementing the large-scale production of a narcotic for a government with strictly enforced monopsony rights – their long-term impacts have broader relevance for the study of colonial legacies. I use local administrative data from the late 19<sup>th</sup> and early 20<sup>th</sup> century to investigate how the presence of a highly lucrative resource altered colonial administrators' incentives and the policies they implemented with respect to the rest of the economy.

My empirical strategy exploits the interaction between two factors that affected an area's likelihood of opium cultivation: geographic suitability and administrative boundaries. The opium agencies were careful to select locations whose conditions were most adapted to cultivation, in terms of the soil and climate, as well as the distance to the factories where raw opium was processed (Richards 1981). However, cultivation was only permitted in certain districts meaning that many suitable locations were left unexploited. Using a regression discontinuity design (RDD), I compare the effect of opium suitability in villages either side of

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<sup>&</sup>lt;sup>2</sup> In the year 1907 for example, the combined total number of cultivators in the Bihar and Benares Agencies was over 1.3 million (Opium Department 1909).

the boundaries that demarcated poppy-growing regions. Under the assumption that, absent opium cultivation, the effects of suitability would be continuous across borders, this approach identifies the causal impact of opium production on present-day outcomes.

This empirical design relies on the ability to predict an area's suitability for poppy cultivation. For a significant part of the opium growing region there exist disaggregated data on the share of agricultural land devoted to opium cultivation at the tehsil (sub-district) level from 1895 to 1905. I combine these data with GIS information on temperature, precipitation, altitude, ruggedness, soil types, soil characteristics (e.g. soil ph), the location of rivers and lakes, and the distance to the factories where raw opium was processed. Given the large number of potential explanatory variables, I use the least absolute shrinkage and selection operator (LASSO)<sup>3</sup> to pick the variables that are the best predictors of opium cultivation and generate an opium suitability index. As this measure is likely to be correlated with production of other crops, I calculate similar suitability indices for other crops as well as an aggregate agricultural suitability index. These measures are used as controls and in placebo tests, to ensure that results are driven by suitability for poppy cultivation.

The second element of the identification strategy are administrative rules that confined opium production to specific areas. In British India, opium cultivation was strictly controlled by local 'sub-agencies' who issued licenses to cultivators in their territory. As the borders of these sub-agencies frequently coincided with colonial district boundaries, it is not reasonable to assume that they only affected long-term development through the likelihood of opium production. Instead, I combine the RDD with a cross-section difference-in-differences approach: comparing villages with different suitability for poppies *within* the same district on both sides of the border. If opium cultivation had significant consequences on long-term development, the difference between suitable and unsuitable villages should be larger within former sub-agencies than in districts where cultivation was prohibited.

I find that a higher likelihood of historical opium cultivation is negatively associated with measures of contemporary village-level development used in the literature (e.g. Banerjee and Iyer 2005, Iyer 2010). In a bandwidth of 20 km from the border, the relative impact of a one standard deviation increase in the opium suitability index in poppy-growing districts is a 3.6% (1.4 percentage point) drop in literacy, a 2.6% (2 percentage point) drop in the likelihood of

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<sup>&</sup>lt;sup>3</sup> Using an alternative machine learning technique (random forest classification) results in a similar selection of predictors and an opium suitability index that is highly correlated with that generated by the LASSO.

having a primary school in the village and a 42.5% (1.36 percentage point) drop in the likelihood of having a healthcare facility in the village. These results are robust to controlling for the environmental suitability for other crops, and to narrowing the bandwidth for the RDD. My results also indicate a negative effect on the likelihood of the village having all-weather road access, but this is only significant at the 10 km bandwidth.

Why should historical opium cultivation have had persistent impacts on human capital and public good provision? One potential mechanism relates to the interests of the local colonial administration which were closely intertwined with the state opium enterprise. In poppy-growing districts – senior British officials formally took on a second role as officers of the opium sub-agency. The literature on the natural resource curse suggests that a dominant commodity that provides a large share of government revenue can undermine the incentives for public good provision (Ross 2003, Caselli and Cunningham 2009). In this context, securing the extraction of opium rents and policing the state's monopsony power were a governmental priority, while the incentive to invest in the wider economy may have been weaker. I provide evidence in support of this mechanism using colonial administrative data. At the district-level, opium production was associated with lower per capita expenditure on education and healthcare. Sub-district-level data shows that localities with a higher suitability for opium had fewer schools and a greater number of policemen stationed in their tehsil. These analyses are constrained by the small sample sizes, but the results are consistent with a focus on security at the expense of human capital and diversified economic development.

This paper contributes to the literature on the long-term consequences of colonial rule. Recent work in this area has focussed on individual sectors in order to isolate the impact of specific colonial policies. Dell (2010) and Lowes and Montero (2017) document persistent negative effects of forced labour – on outcomes including public good provision – in the context of Spanish silver mining in Peru and Belgian rubber concessions in the present-day Democratic Republic of Congo, respectively. By contrast, Dell and Olken (2017) show that the establishment of Dutch sugar factories in Indonesia gave rise to agglomeration effects, meaning that nearby areas forced to grow sugar, are now more industrialised, better educated and have better infrastructure. The contrast between sugar and opium is instructive. While the former was a widely consumed commodity with important downstream linkages to food processing and manufacturing, opium in India was cut off from the rest of the economy – by virtue of its distinctive uses but also deliberate British policy. When the Dutch sugar factories closed, the economic opportunities that had sprung up around them remained. The decline of opium

production in India may be more reminiscent of the depletion of a mineral resource, with few enduring benefits.

My paper also contributes to a strand of this literature which studies the persistent effects of different forms of colonial rule in India. Banerjee and Iyer (2005) show that landlord-based tenure systems perpetuated by the British in some regions, led to lower agricultural investment and productivity after independence. Iyer (2010) exploits a colonial annexation rule to compare the effect of direct and indirect rule on long-run development, finding that areas formerly under indirect rule have better access to schools, health centres and roads. The study of opium is interesting in this context, because there is variation in cultivation both within British India and between princely states. In section 6, I analyse the effect of opium suitability on literacy and public good provision in princely states and find no significant negative effect. As I cannot directly replicate my empirical strategy in this sample, this difference should be interpreted with caution. One potential explanation, however, is that the private merchants who were free to enter the trade in princely states were more likely to invest their profits locally than the EIC or the British Government.

The remainder of this paper is structured as follows. Section 2 provides background on the history of opium in India and the institutions the British implemented to administer production. Section 3 describes the data used in the analysis and the construction of the agricultural suitability indices. Section 4 explains the empirical strategy and tests the validity of assumptions required for identification. The results are presented in Section 5. Section 6 provides evidence on potential mechanisms for the persistent effects of opium cultivation. Section 7 concludes.

# 2. Background

#### 2.1 Opium cultivation in India

The cultivation of opium in India predates the arrival of European colonial powers. It is thought to have been first introduced by Arab traders in the 8<sup>th</sup> century (Watt 1892). Opium was established as a lucrative cash crop in the 17<sup>th</sup> and 18<sup>th</sup> centuries as Dutch traders, and later the EIC, bought up increasing quantities from Indian merchants in order to meet growing demand in China. After cementing its political control over Bengal, the EIC was able to declare a state

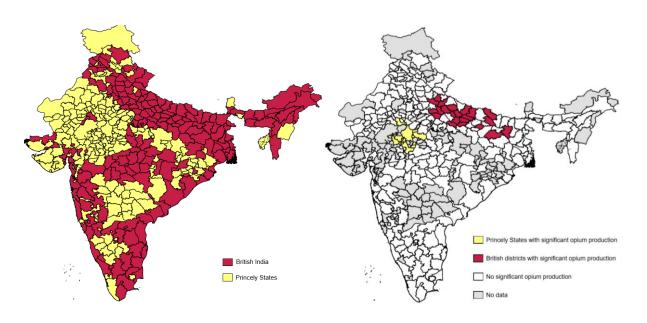
monopoly on the cultivation and sale of the crop in 1772. Procurement of raw opium was initially the purview of British contractors whose abuse of their official powers made them "notorious" in cultivating areas (Richards 1981). Farmers were often forcibly coerced into growing opium rather than traditional staple crops (Wright 1959), a practice which exacerbated famines in times of crop failure (Chaudhury 2003, Ul Haq 2000). Remuneration was not always adequate as contractors were known to force cultivators to give up their crop for less than the cost of production. To regulate cultivation, the EIC set up official government agencies in 1797 (Sahu 1977). This 'agency system' remained in place until the decline of the opium trade following the Hague International Opium Convention in 1912.

The two opium agencies of Bihar and Benares organised cultivation in large parts of the Gangetic plain, in the present-day states of Uttar Pradesh, Bihar and Jharkhand. They were divided into sub-agencies (11 for Bihar and 16 for the Benares agency). Outside the borders of these sub-agencies cultivation was prohibited throughout British India. Each sub-agency was led by a British official who commanded a network of Indian agents (known as *gumashtas*). The gumashtas selected areas suitable for cultivation and negotiated with village headmen before issuing licenses to cultivators (Richards 1981). No farmer was allowed to grow opium without a license or to sell their crop to anyone other than the opium agency. Cultivators received an advance that would be deducted from their final payment upon delivery of the harvest. Farmers unable to meet their production quota fell into debt to the agency which could be paid off in the form of future opium harvests (Allen 1853). Once harvested, the raw opium was shipped to one of two central factories - at Ghazipur (for the Benares agency) or Patna (for the Bihar agency) — where it was processed and packed into chests for export.

The EIC – and later the British government – were only able to enforce their monopsony inside areas under direct British rule (see Map 1). Given the profitability of opium, various princely states in the Malwa region of western India emerged as competitors in the early 19<sup>th</sup> century (see Map 2). The British initially sought to curtail this trade by forbidding exports of Malwa opium from the port of Bombay, then to control it by seeking to buy up the entire output, before finally allowing the trade while charging a high excise fee on shipments passing through Bombay. In section 6, I analyse whether opium cultivation is associated with negative long-run effects in princely states, in order to distinguish between the effect of the crop per se and the colonial institutions set up to regulate its cultivation in directly-ruled areas.

Map 1: Direct and indirect rule in India 1907

**Map 2: Opium cultivating areas** 



Note: Map 1 depicts the districts of British India in red and princely states in yellow. The source is the Imperial Gazetteer of India 1907. Map 2 shows areas where opium cultivation accounted for at least 1% of total agricultural acreage on average between 1891 and 1911 (source: Agricultural Statistics of India). Areas shaded in grey are princely states for which no data is available.

## 2.2 Opium and local development

The consequences of the opium trade were the subject of intense political debate in Britain and India throughout the 19<sup>th</sup> century. In part, this debate centred on the morality of exporting a narcotic to China against the wishes of its government. However, many commentators also suggested that the trade had adverse consequences on the population in cultivating areas.

The agency system ensured that farmers did not share in the large profits of the opium trade.<sup>4</sup> Given their monopsony power, the opium agencies were able to "keep the price of crude opium just on the economic edge" (Richards 1981). It was common for cultivators to protest that the price was too low for production that required fertile soil, irrigation, and intensive labour input (Royal Commission on Opium 1894). British officials argued that poppy cultivation was advantageous due to its lack of risk; farmers faced a dependable buyer that operated in the same region for decades. They also suggested that the advances paid before the growing season allowed for investments in wells and irrigation (in section 5 I test whether opium cultivation

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<sup>&</sup>lt;sup>4</sup> In 1881 for example, the cost to the Indian government of the raw opium for one chest was estimated at Rs 370. This chest could then be sold to exporters in Calcutta for Rs 1362 (Richards 2002), giving the Opium Agency a 73% profit margin.

had any lasting influence on irrigation infrastructure). In terms of long-run agricultural development, specialising in poppies may have come at the expense of developing expertise in other crops. By the time opium cultivation dropped sharply following the Opium Convention in 1912, areas where growing poppies had been prohibited may have developed comparative advantages in other crops. Moreover, other crops may have exhibited input-output linkages leading to growth in sectors such as food processing or textiles. The agency system ensured that poppy growing was cut off from the rest of the economy, with the only downstream activity confined to the two government factories that processed the raw opium. In this respect, poppy production in British India was very different from the forced sugar cultivation in Indonesia which Dell and Olken (2017) argue had long-term benefits through agglomeration effects.

A separate set of potential developmental consequences relate to opium's role as a narcotic. In addition to its monopsony on raw opium, the British government had a monopoly on the sale of opium in India. While the majority of processed opium was exported, some was sold domestically through licensed shops. In poppy-growing areas the authorities could not control the population's access to unprocessed opium and contemporary sources report widespread consumption in some regions (Royal Commission on Opium 1894). The strict price controls also gave rise to organised smuggling which became a significant preoccupation for colonial authorities. The government invested in a network of spies and informants and conducted antismuggling drives which "often ended up harassing honest poppy cultivators" (Deshpande 2009). It is possible that poppy-growing areas suffered directly from the availability of a crop that enabled drug use and crime.

Opium production may also have had persistent effects by changing the incentives of colonial officials. Each district of British India was governed by a district collector. In districts that formed part of opium sub-agencies, these collectors automatically also became officers of the opium agency with the rank of Deputy Opium Agent. While it is not clear from historical sources whether this dual role proved a constraint in terms of fulfilling the normal administrative functions of a collector, it may have affected their priorities. The literature on the natural resource curse suggests that the presence of a commodity that generates large rents and is not integrated in the wider economy can have a detrimental impact on governance (see for example Ross 1999, Sala-i-Martin and Subramanian 2013, Frankel 2010). Opium was similar to a mineral resource in that it generated very significant revenues while employing a comparatively small share of economic inputs in terms of labour, capital or land. The share of

agricultural land in British India devoted to opium was 0.29% at its peak, but the contribution to total colonial revenue could be as high as 15%. In opium-growing districts the average share of agricultural land was only around 3% while the share of local economic output would be many times higher. In this context, local officials were likely to have a large incentive to guarantee the stable extraction of opium rents and to police the state's monopsony and monopoly powers. By contrast, they may have had less interest in investing in the wider local economy, through infrastructure, education or healthcare provision than their counterparts in districts with no opium production and no personal affiliation to the opium agencies. In section 6, I test this mechanism by evaluating the relationship between opium production and the strength of local police forces at the time (a measure of expenditure on security) as well as the number of local schools (a measure of human capital investment).

## 3. Data

This project combines colonial administrative records on agricultural production and public services with modern census data as well as GIS information on climactic conditions. These data at available at different levels of aggregation. Historical sources provide data either at the level of the colonial district and princely state, or in some cases, at the level of the tehsil (sub-district). Contemporary census data is at the village level. The GIS data used for this project were initially at the level of a 30 arc second grid (~1 km<sup>2</sup>).

#### 3.1 Historical data on opium cultivation

The primary source of data on opium cultivation for this study are the District Gazetteers of the United Provinces of Agra and Oudh (1909). These provide comparatively disaggregated agricultural statistics at the tehsil level for the period 1895 to 1905. Each district's gazetteer has a separate table for each tehsil with an annual breakdown of acreage allocated to all of the major crops including opium. I digitised these records and calculated the average share of agricultural acreage devoted to opium for each tehsil. The provinces of Agra and Oudh included

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<sup>&</sup>lt;sup>5</sup> Given that British-controlled opium production started in the late-18<sup>th</sup> century and lasted for over a century, it is important to consider to what extent the statistics available for a comparatively short period are representative. The total acreage devoted to opium increased progressively until the 1870s and then remained comparatively stable until 1905. The administrative structure of the opium agencies and the technology of production were similarly stable over this latter period. The data can therefore be considered representative of the last 40 years of opium production – a time when the area under cultivation had reached its greatest extent.

the entire Benares Opium Agency which accounted for roughly 70% of British-controlled opium production during this period.

The Agricultural Statistics of India provide a less detailed but more comprehensive coverage of opium production from 1891 to 1911. These data allow me to calculate the equivalent share of agricultural acreage assigned to opium for all districts in British India as well as for 96 territories in former princely states.

#### 3.2 Climate and soil data

My identification strategy relies on fine-grained variation in local climatic conditions. Fick and Hijmans (2017) provide 30 arc second grid-level data on a range of bioclimatic variables including monthly measures of temperature and precipitation. The DIVA-GIS database of Hijmans et al. (2012) is the source for data on altitude and the location of rivers and lakes. I use the measure of terrain ruggedness from Shaver et al. (2016). Data on local soil composition and characteristics is from the International Soil Reference and Information Centre (ISRIC). Based on the soil's content of sand, silt, and clay, I assign it to one of twelve basic soil types using the USDA soil classification system. The full list of environmental variables is provided in panel C of Table 1.

#### 3.3 Opium suitability index

There is no existing global crop suitability index for the opium poppy (*papaver somniferum*).<sup>6</sup> I construct an index using the tehsil-level data on opium cultivation from the provinces of Agra and Oudh and the climactic and soil variables discussed in the previous sub-section. All geographic variables are aggregated to the tehsil-level by taking the mean of the grid cells inside the tehsil borders. The resulting sample has 216 observations. Given that the number of potential predictors is large relative to the number of tehsils, I use the machine-learning technique LASSO to identify a subset of variables that explain the greatest share of variation in opium cultivation. To avoid overfitting, I use 10-fold cross-validation to select the optimal lambda – the shrinkage parameter which penalizes the coefficients on potential predictors.

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<sup>&</sup>lt;sup>6</sup> Kienberger et al. (2017) conduct a spatial risk assessment for opium cultivation in Afghanistan which includes a measure of environmental suitability as well as socio-economic vulnerability. The construction of their environmental suitability index involves land use data, which would not be appropriate for this study as it is likely to be endogenous to historical agricultural activity.

Following Belloni and Chernozhukov (2013), I re-estimate the model selected by LASSO in OLS to obtain the coefficients for the index.

The model selected includes nine variables: the distance to water, the minimum temperature in the coldest month of the year, precipitation seasonality (the coefficient of variation in monthly precipitation), the organic carbon content of the soil, the ph of the soil, the sand content of the soil, the share of soil of the sandy-clay-loam class, the share of soil of the clay-loam class and the distance to the opium factory (at Ghazipur for villages in the Benares agency and Patna for villages in the Bihar agency). The climactic and soil variables are in line with historical sources which suggest that soil of the clay-loam type and proximity to water provided favourable conditions for cultivation (Richards 1981). It is also intuitive that the distance to the opium factory improves the model fit. All raw opium had to be shipped to these factories for processing so, within each district, selecting areas closer to the factory resulted in lower transport costs. The locations of the two factories were not randomly selected (prior to the arrival of the EIC there had been historical opium exchanges at Ghazipur and Patna) but could be considered exogenous to the fine-grained local variation around district borders exploited in the estimation. The average village in the sample is 246 km away from the factory, while the bandwidth for the RDD is 20 or 10km. Nevertheless, the results are robust to the use of an opium suitability index that omits the distance to the factory.

To compute village-level opium suitability, I construct a 5 by 5km grid and estimate the value of the suitability index for each grid cell. The validity of the opium suitability index and its out-of-sample performance are assessed in section 4. I also replicate this approach to construct an aggregate agricultural suitability index (based on the share of available land that is cultivated), as well as suitability indices for wheat, cotton, sugarcane, tea, indigo and tobacco. All suitability indices are standardised to have a mean of 0 and a standard deviation of 1.

#### 3.4 Census data

The dependent variables are drawn from the 2011 Population Census of India. While this source does not have a direct measure of village-level income it includes a number of variables that have been used as proxies for economic development in the literature (Banerjee and Iyer 2005, Iyer 2010). The Village Amenities module of the 2011 census contains information on a wide range of public goods. I use dummy variables for whether a village has a primary school,

a health centre, and all-weather road access. I also include literacy as a measure of village-level human capital. To test the theory that opium cultivation may have had lasting impacts on irrigation infrastructure, I use the share of agricultural land that is irrigated and the share that is irrigated by wells.

#### 3.5 Colonial administrative data

In section 6, I analyse the relationship between opium cultivation and the allocation of public resources by British officials. For this purpose, I digitise data on public administration from the District Gazetteers of Agra and Oudh (1909). These record the annual breakdown of local fiscal spending across different categories including expenditure on health and education. The gazetteers also provide a comprehensive list of schools and police stations (and the number of officers in each station). I map these to the tehsil-level in order to evaluate the relative emphasis on security and human capital in opium growing areas.

#### 3.6 Sample and descriptive statistics

Table 1 presents descriptive statistics for the variables used in the analysis. The main sample is drawn from 190,318 villages in the 2011 census that are within the borders of the former opium-growing provinces of Agra, Oudh and Bengal. For the RDD analysis this sample is restricted to villages within 20 km or 10 km of the borders that demarcated opium sub-agencies, giving a sample of 24,659 and 13,489 villages respectively. About half of the villages (52%) were located in districts that formed part of sub-agencies, where opium cultivation was permitted. The share of agricultural land devoted to opium was comparatively small: 2% at the tehsil level in the provinces and Agra and Oudh (panel D). However, this share could be significantly higher within opium sub-agencies, reaching a maximum of 43% at the tehsil-level. This distribution is convenient for my empirical analysis in that the effects of opium production are likely to be concentrated in locations particularly favourable for cultivation.

**Table 1: Descriptive Statistics** 

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Variable	Obs	Mean	Std. Dev.	Min	Max
Panel A: village characteristics					
Population	190318	1631.33	2092.31	1.00	43722.00
Village area	190313	213.33	396.26	0.00	88977.82
literacy	190318	0.45	0.16	0.00	1.00
primary school	190318	0.78	0.42	0.00	1.00
all-weather road	190318	0.64	0.48	0.00	1.00
health centre	190318	0.03	0.18	0.00	1.00
share of agr. area irrigated	186333	0.70	0.35	0.00	1.00
share irrigated by wells	186824	0.35	0.39	0.00	1.00
distance to Delhi (km)	190318	722.74	373.87	15.27	1397.96
distance to Calcutta (km)	190318	651.77	358.18	9.08	1399.15
Panel B: administrative status and sui					
opiumdistrict	190318	0.52	0.50	0.00	1.00
suitability	190315	-0.09	1.01	-8.80	5.31
aggregate agr. Suitability	190315	0.11	0.97	-4.31	5.78
wheat suitability	190318	0.02	1.07	-11.56	0.99
distance (to border)	186161	111.32	90.29	0.00	447.05
distance *opiumdistrict	186161	40.05	51.64	0.00	208.30
opiumdistrict*suitability	190315	0.03	0.55	-3.19	4.96
distance*suitability	186158	29.52	158.97	-1287.22	2691.94
distance *opiumdistrict*suitability	186158	-0.59	43.68	-343.20	907.93
Panel C: geographic variables for LA					
ruggedness	190318	45.21	140.12	0.00	1477.89
altitude	190318	182.98	335.88	0.62	4414.19
distance to water	190318	13.67	14.00	0.00	147.75
distance to Ghazipur factory	190318	359.26	206.67	0.73	819.31
distance to Patna factory	190318	392.21	221.68	2.85	931.46
annual mean temperature	190318	252.00	22.89	12.28	275.97
mean diurnal range	190318	116.71	12.53	65.94	141.94
isothermality	190318	40.15	1.77	34.03	47.19
temperature seasonality	190318	5240.88	901.36	2667.23	6702.97
max temp. warmest month	190318	386.19	32.64	135.92	430.75
min temp. coldes month	190318	97.82	25.90	-128.03	170.97
temp. annual range	190318	288.37	38.02	155.51	349.03
mean temp. wettest quarter	190318	285.03	20.59	76.83	300.31
mean temp. driest quarter	190318	228.68	44.20	-24.03	307.14
mean temp. warmest quarter	190318	309.40	25.30	83.22	339.11
mean temp. coldest quarter	190318	176.13	26.88	-65.72	230.28
annual precipitation	190318	1179.82	271.37	599.89	3591.06
precipitation wettest month	190318	329.83	64.72	195.06	976.31
precipitation driest month	190318	3.00	2.25	0.00	30.25
precipitation direct month	190318	119.46	14.60	65.25	154.14
precipitation seasonanty precipitation wettest quarter	190318	837.19	147.79	499.89	2435.53
precipitation driest quarter	190318	30.96	19.39	4.86	150.00
precipitation warmest quarter	190318	340.88	222.59	39.00	2328.75
precipitation coldest quarter	190318	43.47	29.83	13.06	312.83
soil clay content	190315	31.66	3.89	10.20	49.88
son clay content	170513	31.00	3.07	10.20	77.00

soil ph	190315	55.34	3.17	43.81	69.56
soil silt content	190315	34.72	4.97	20.03	49.68
soil sand content	190315	33.62	5.67	15.13	58.60
soil organic carbon content	190315	3.97	4.30	0.00	55.82
soil type sand	190318	0.00	0.00	0.00	0.00
soil type loamy sand	190318	0.00	0.00	0.00	0.00
soil type sandy loam	190318	0.00	0.02	0.00	1.00
soil type silt loam	190318	0.00	0.00	0.00	0.00
soil type silt	190318	0.00	0.00	0.00	0.00
soil type silt clay loam	190318	0.01	0.10	0.00	1.00
soil type sand clay	190318	0.00	0.00	0.00	1.00
soil type silty clay	190318	0.00	0.00	0.00	1.00
soil type clay	190318	0.01	0.12	0.00	1.00
Panel D: Tehsil-level data for Agra a	ınd Oudh 1895-	1905			
share of total land area cultivated	216	0.61	0.14	0.21	0.90
opium share of cultivated land	216	0.02	0.04	0.00	0.43
wheat share of cultivated land	216	0.32	0.10	0.06	0.74
Panel E: District and princely state l	evel data 1891-	1911			
british district	382	0.47	0.50	0.00	1.00
princely state	382	0.52	0.50	0.00	1.00
total land area cultivated	262	757519	646348	3596	3311048
indigo share of cultivated land	262	0.00	0.00	0.00	0.03
opium share of cultivated land	262	0.00	0.01	0.00	0.06
tea share of cultivated land	262	0.01	0.03	0.00	0.30
tobacco share of cultivated land	262	0.00	0.01	0.00	0.13
sugarcane share of cultivated land	262	0.02	0.04	0.00	0.24
cotton share of cultivated land	262	0.04	0.09	0.00	0.50

Note: Panels A to C presents descriptive statistics for villages in the former opium growing provinces of Agra, Oudh and Bengal. Geographic variables and suitability indices are calculated for a 5 by 5 km grid, and villages are assigned the values corresponding to their grid cell. Panel D presents sub-district level data from the district gazetteers of the United provinces of Agra and Oudh. Panel E presents district or princely state-level data for the whole of India.

# 4. Empirical Strategy

#### 4.1 Empirical strategy

The location of opium sub-agencies was not randomly assigned and is likely to be correlated with unobservable factors that affect present-day development outcomes. In order to identify the causal effect of opium production, I combine a regression discontinuity (RD) design with a cross-section difference-in-differences (DiD) approach. The intuition behind this empirical strategy is that geographic suitability for poppies should have increased the likelihood of cultivation in opium-growing districts while having no such effect in districts where cultivation was prohibited. I restrict the sample to villages close to the historical borders that demarcated

poppy cultivation, control flexibly for the distance to these borders, and include district fixed effects. In this framework, I evaluate the impact of opium cultivation on contemporary development by testing for a discontinuity in the effect of opium suitability across the border. This section describes the estimation strategy before discussing the critical identification assumption: that absent opium cultivation, the effect of suitability would be continuous across district borders.

$$Y_{id} = \alpha + \beta_1 suitability_{id} + \beta_2 suitability * opiumdistrict_{id} + \mathbf{D}_{id} + \mathbf{X}_{id} + \delta_d + \varepsilon_{id}$$
 (1)  
 $\forall i \ where \ distance_{id} \in [-\mu, \mu]$ 

The main results are the product of estimating equation (1). The dependent variables  $Y_{id}$  are measures of village-level development from the 2011 Population Census – literacy, primary school access, health centre access, and all-weather road access – for village i in colonial district d. opiumdistrictid is a dummy variable that takes the value of 1 if a village is situated within the borders of a colonial district where opium was cultivated. The coefficient of interest is  $\beta_2$  which captures the differential effect of poppy suitability in opium-growing districts.

Equation (1) corresponds to a local linear RD estimation. For each village  $distance_{id}$  gives the distance to the closest border between an opium-growing district and a district where cultivation was prohibited. This is the running variable in the RD. It is a continuous measure across the border, taking positive values where  $opiumdistrict_{id} = 1$  and negative values where  $opiumdistrict_{id} = 0$ . The vector  $\mathbf{D}_{id}$  includes  $distance_{id}$  as well as its interactions with  $opiumdistrict_{id}$ ,  $suitability_{id}$ , and  $suitability*opiumdistrict_{id}$ . These are included in order to flexibly control for the effect of distance to the border depending on district status and poppy suitability.  $\mathbf{X}_{id}$  is a vector of additional geographic controls. Colonial district fixed effects  $\delta_d$  ensure that the results are driven by within-district variation in opium suitability. Section 5 presents results for two bandwidths  $\mu$ : 10km and 20km.

#### 4.2 Identifying assumptions

My empirical strategy relies on three assumptions which can be seen as analogous to those underlying instrumental variable estimation. First, I assume that my measure of poppy

<sup>&</sup>lt;sup>7</sup> This variable is not included directly in the regression equation as it is colinear with the colonial district fixed effects  $\delta_d$ 

suitability is correlated with historical opium cultivation. Second, I assume that there was a discontinuity in the likelihood of cultivation at district borders, i.e. that colonial rules were enforced. These assumptions are comparable to the relevance criterion for an instrumental variable. Finally, identification relies on the assumption that poppy suitability would only have a discontinuous effect across historical opium agency borders as a result of opium production. This is analogous to the exclusion restriction for an instrumental variable.

In the absence of village-level data on opium cultivation I evaluate the first two assumptions using the two historical datasets discussed in section 3: tehsil-level data for the United Provinces of Agra and Oudh for the period 1895-1905 and district-level data for British India and the princely states for the period 1891-1911. Table 2 presents an evaluation of the relationship between poppy suitability and the share of acreage devoted to opium. In the tehsil-level sample the first assumption holds more or less by construction, as this data was used to select the best predictors of cultivation. The F-statistic in a regression of opium acreage share on suitability is 76.5 and the RMSE is 0.034. However, the subsequent rows show that the suitability index is also correlated with observed cultivation out-of-sample and at different levels of aggregation. <sup>8</sup> The F-statistic is consistently well above 10 and the RMSE is relatively stable across samples albeit slightly higher for princely states. <sup>9</sup>

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<sup>&</sup>lt;sup>8</sup> A remaining concern is, that the relationships observed at the tehsil- and district- level may not hold at the village level. While all of the geographic inputs to the suitability index should be scalable to any geographic unit and there is considerable variation in the size of the tehsils and districts in the samples for Table 2, the relationship with opium production may be conditioned by administrative factors. For this reason, the last row of Table 2 evaluates the predictive performance in geographic areas that are randomly generated configurations of tehsils that do not correspond to any historical administrative units. This does not reduce the predictive performance of the suitability index.

<sup>&</sup>lt;sup>9</sup> A difficulty that arises when applying the same opium suitability index is, that one component – the distance to the opium factory – has no direct equivalent in indirectly ruled areas. At present, I have no data on the locations where opium was processed in princely states. As such, I use the version of the opium suitability index that excludes distance to the factory and also present results for a version where the distance to the port of Bombay (the central location through which Malwa opium passed before export) is included instead. The latter performs slightly better in terms of explanatory power and goodness of fit.

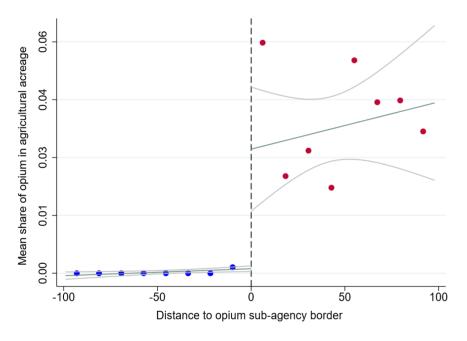
Table 2: Validity of opium suitability index

Sample	Level	N	F-statistic	RMSE
Agra and Oudh	tehsil	216	76.51	0.034
Agra and Oudh	district	48	34.56	0.018
Bengal	district	32	31.48	0.025
British India	district	164	32.91	0.048
Princely states (no distance)	state	88	23.83	0.054
Princely states (distance to port)	state	88	26.27	0.045
Inter-district tehsil configurations	tehsil	40976	15227.46	0.024

Note: This table evaluates the opium suitability index's explanatory and predictive power for actual historical data on opium's share of agricultural acreage in different samples. The first row corresponds to the data used for the construction of the index. See footnotes 8 and 9 for additional explanations on the 7<sup>th</sup>, 5<sup>th</sup> and 6<sup>th</sup> rows.

Figure 1 plots a graphical test of the second assumption – that colonial rules on opium cultivation were enforced. There is a sharp discontinuity in the share of acreage devoted to opium at the border to opium-growing districts. Tehsil-level data shows production was zero or very close to zero in all districts where cultivation was prohibited, even those bordering opium sub-agencies. These areas can therefore be considered an uncontaminated control group in the empirical analysis.

Figure 1: Testing for a discontinuity in production at opium sub-agency borders



Note: Chart plots the share of opium in agricultural production either side of the borders that demarcated cultivating areas. Tehsils are grouped into bins based on their proximity to the border. The left side of the cut-off corresponds to the areas where opium cultivation was prohibited.

The estimation strategy outlined above will provide a causal estimate of the effect of opium production on contemporary outcomes provided that there are no unrelated factors that would cause poppy suitability to have differential effects across opium agency borders. I conduct several tests to evaluate the validity of this final assumption. First, I test for discontinuities in a range of geographic variables at the border (Table 3). Out of 36 variables only one – the share of soil class clay – exhibits a discontinuity at the 90% significance level. This imbalance is not inconsistent with chance and I control for the imbalanced variable in subsequent regressions. Importantly, there is no discontinuity in geographic suitability for poppy cultivation or aggregate agricultural suitability at former agency borders. Second, I conduct placebo tests which replicate the estimation of equation (1) but substitute poppy suitability with suitability for other crops. These crops were not restricted to opium agency borders and so a discontinuous effect on literacy and public good availability would call the identification strategy into question. Table 4 shows that no crop exhibits a systematic discontinuity <sup>10</sup>, which is supporting evidence that the results in the next section are driven by opium production.

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<sup>&</sup>lt;sup>10</sup> Two regressions yield a significant coefficient on the interaction term of interest (health centre on indigo suitability, and literacy on sugarcane). This number of significant coefficients is not inconsistent with chance and both suitability indices have no consistent positive or negative effect across other dependent variables.

Table 3: Test for discontinuities at opium-sub agency borders

Table 3: Test for discontinui	Coefficient on	agency borders
Dependent variable	opiumdistrict	Standard error
suitability	0.018	(0.061)
aggregate agricultural suitability	0.063	(0.051)
wheat suitability	0.003	(0.032) $(0.019)$
ruggedness	0.001	(1.853)
altitude	1.972	(7.040)
distance to water	0.342	(0.761)
distance to Ghazipur factory	16.285	(11.703)
distance to Patna factory	15.283	(13.936)
•	-0.118	(0.335)
annual mean temperature mean diurnal range	0.116	(0.542)
_	0.003	(0.342) $(0.063)$
isothermality		
temperature seasonality	15.250	(41.180)
max temp. warmest month	-0.154	(1.094)
min temp. coldest month	-0.401	(0.787)
temp. annual range	0.247	(1.620)
mean temp. wettest quarter	0.189	(0.475)
mean temp. driest quarter	-0.214	(2.886)
mean temp. warmest quarter	0.045	(0.649)
mean temp. coldest quarter	-0.348	(0.658)
annual precipitation	-5.729	(9.072)
precipitation wettest month	-3.963	(2.837)
precipitation driest month	0.007	(0.066)
precipitation seasonality	-0.379	(0.682)
precipitation wettest quarter	-6.243	(5.653)
precipitation driest quarter	0.157	(0.380)
precipitation warmest quarter	4.919	(9.445)
precipitation coldest quarter	-0.085	(0.620)
soil clay content	-0.12	(0.267)
soil ph	0.213	(0.234)
soil silt content	0.241	(0.399)
soil sand content	-0.120	(0.342)
soil organic carbon content	0.103	(0.112)
soil type clay	-0.0125*	(0.007)
soil type loam	-0.004	(0.019)
soil type sandy clay loam	0.0165	(0.016)
soil type clay loam	-0.001	(0.025)
N	339	911
Bandwidth	20	km
Controls used in all regressions	distance, distance	ce*opiumdistrict

Note: Each row corresponds to an RDD regression of the respective variable on opiumdistrict, distance and distance\*opiumdistrict within the 20km bandwidth. Some soil types listed in Table 1 are omitted as they do not occur within the sample. \*\*\* p<0.01, \*\* p<0.05, \* p<0.1

Table 4: Placebo tests - interaction between subagency borders and crop suitability

Variable	literacy	primary school	health centre	road access
wheat				
suitability*opiumdistrict	-0.0326	-0.0206	-0.0195	0.0062
	(0.0242)	(0.0203)	(0.0251)	(0.0059)
indigo				
suitability*opiumdistrict	0.0031	-0.0008	-0.0005*	-0.0010
	(0.0057)	(0.0010)	(0.0003)	(0.0013)
tea				
suitability*opiumdistrict	0.0007	0.0005	0.0007	-0.0002
	(0.0007)	(0.0009)	(0.0012)	(0.0011)
tobacco				
suitability*opiumdistrict	0.0031	-0.0049	0.0156	0.0127
	(0.0096)	(0.0034)	(0.0112)	(0.0102)
sugarcane				
suitability*opiumdistrict	0.0136**	0.0115	0.0035	-0.0139
	(0.0062)	(0.0141)	(0.0059)	(0.0141)
cotton				
suitability*opiumdistrict	0.0044	0.0006	0.0003	-0.0090
	(0.0074)	(0.0011)	(0.0011)	(0.0061)
aggregate agricultural				
suitability*opiumdistrict	-0.0017	0.0053	0.0073	-0.0088
	(0.0051)	(0.0165)	(0.0074)	(0.0184)

N 24,659 Bandwidth 20 km District fixed effects Yes

 $Controls:\ crop\ suitability, distance, crop\ suitability* distance, distance* opium district,$ 

crop suitability\*distance\*opiumdistrict

Note: Table reports results from 28 separate regressions. These correspond to estimating equation (1) for 4 dependent variables, replacing opium suitability with 7 suitability indices. \*\*\* p<0.01, \*\* p<0.05, \* p<0.1

# 5. Results

Table 5 presents the results of estimating equation (1) in an RDD with bandwidths of 20 km and 10 km. For literacy, primary school access, and health centre access the coefficient on the interaction between opiumdistrict and opium suitability is negative and significant. The magnitude of the coefficient is stable across bandwidths and robust to controlling for suitability to other crops. At the 20 km bandwidth with controls, the relative impact of a one standard deviation increase in the opium suitability index in poppy-growing districts is a 1.4 percentage point drop in literacy, a 2 percentage point drop in the likelihood of primary school access and a 1.36 percentage point drop in the likelihood of having a healthcare facility in the village. Relative to the mean dependent variable in non-poppy growing districts, these estimates correspond to declines of 3.6% for literacy, 2.6% for primary school access and 42.5% for health centre access (where the control mean is only 3.2%)

Table 5A: RD estimates at the 20 km bandwidth

Lite	racy	Primary school		Health centre		All-weather road access	
(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
-0.0151***	-0.0143***	-0.0209**	-0.0207**	-0.0144***	-0.0136**	-0.0139	-0.0120
(0.00283)	(0.00280)	(0.00834)	(0.00846)	(0.00540)	(0.00546)	(0.00997)	(0.0100)
0.0021	0.0022	0.0109*	0.0020	0.00195	0.00222	-0.00150	-0.00831
(0.0022)	(0.0022)	(0.00562)	(0.0058)	(0.00452)	(0.00457)	(0.00719)	(0.00725)
	0.0214***		0.00718		-0.000201		0.0370***
	(0.00220)		(0.00715)		(0.00324)		(0.00781)
	0.119***		0.0411		0.0106		0.0269
	(0.00890)		(0.0277)		(0.00986)		(0.0281)
	-0.129***		-0.122		0.0884		-0.328***
	(0.0319)		(0.124)		(0.0844)		(0.113)
	0.0947***		0.221***		0.0536**		-0.0913*
	(0.0161)		(0.0510)		(0.0268)		(0.0522)
	0.0607***		0.591***		0.109***		0.241***
	(0.0196)		(0.0573)		(0.0268)		(0.0687)
	0.00197**		-0.0241***		0.00133		-0.0129***
	(0.000891)		(0.00288)		(0.00130)		(0.00339)
24,659	24,388	24,659	24,388	24,659	24,388	24,659	24,388
X	X	X	X	X	X	X	X
X	X	X	X	X	X	X	X
0.289	0.308	0.100	0.109	0.007	0.009	0.063	0.069
0.398	0.399	0.749	0.748	0.029	0.029	0.728	0.728
	(1) -0.0151*** (0.00283) 0.0021 (0.0022) 24,659 X X 0.289	-0.0151*** -0.0143*** (0.00283) (0.00280) 0.0021 0.0022 (0.0022) (0.0022) 0.0214*** (0.00220) 0.119*** (0.00890) -0.129*** (0.0319) 0.0947*** (0.0161) 0.0607*** (0.0196) 0.00197** (0.000891)  24,659 24,388 X X X X X X X X O.289 0.308	(1) (2) (3) -0.0151*** -0.0143*** -0.0209** (0.00283) (0.00280) (0.00834) 0.0021 0.0022 0.0109* (0.0022) (0.0022) (0.00562) 0.0214*** (0.00220) 0.119*** (0.00890) -0.129*** (0.0161) 0.0607*** (0.0196) 0.00197** (0.000891)  24,659 24,388 24,659 X X X X X X X X X O.289 0.308 0.100	(1)         (2)         (3)         (4)           -0.0151***         -0.0143***         -0.0209**         -0.0207**           (0.00283)         (0.00280)         (0.00834)         (0.00846)           0.0021         0.0022         0.0109*         0.0020           (0.0022)         (0.00562)         (0.0058)           0.0214***         0.00718         (0.00715)           0.119***         0.0411         (0.00890)           (0.0319)         (0.124)         (0.0277)           -0.129***         -0.122         (0.0319)         (0.124)           0.0947***         0.221****         (0.0510)           0.0607***         0.591***         (0.0573)           0.00197**         -0.0241***         -0.0241***           (0.000891)         (0.00288)           24,659         24,388         24,659         24,388           X         X         X         X           X         X         X         X           X         X         X         X           X         X         X         X           X         X         X         X           X         X         X         X	(1)         (2)         (3)         (4)         (5)           -0.0151***         -0.0143***         -0.0209**         -0.0207**         -0.0144***           (0.00283)         (0.00280)         (0.00834)         (0.00846)         (0.00540)           0.0021         0.0022         0.0109*         0.0020         0.00195           (0.0022)         (0.00562)         (0.0058)         (0.00452)           0.0214***         0.00718         (0.00715)           0.119***         0.0411         (0.0277)           -0.129***         -0.122         (0.0319)         (0.124)           0.0947***         0.221***         (0.0161)         (0.0570)           0.0607***         0.591***         (0.0041***           (0.009891)         (0.00288)         0.00241***           24,659         24,388         24,659         24,388         24,659           X         X         X         X         X         X           X         X         X         X         X         X           X         X         X         X         X         X           X         X         X         X         X         X           X	(1)         (2)         (3)         (4)         (5)         (6)           -0.0151***         -0.0143***         -0.0209**         -0.0207**         -0.0144***         -0.0136**           (0.00283)         (0.00280)         (0.00834)         (0.00846)         (0.00540)         (0.00546)           0.0021         0.0022         0.0109*         0.0020         0.00195         0.00222           (0.0022)         (0.0022)         (0.00562)         (0.0058)         (0.00452)         (0.00457)           0.0214***         0.00718         -0.000201         (0.00324)         (0.00715)         (0.00324)           0.119***         0.0411         0.0106         (0.00890)         (0.0277)         (0.00986)           -0.129***         -0.122         0.0884         (0.0319)         (0.124)         (0.0844)           0.0947***         0.221***         0.0536**         (0.0161)         (0.0510)         (0.0268)           0.0607***         0.591***         0.109***         (0.0268)           0.00197**         -0.0241***         0.00133           (0.000891)         (0.00288)         (0.00130)           24,659         24,388         24,659         24,388           X         X	(1)         (2)         (3)         (4)         (5)         (6)         (7)           -0.0151***         -0.0143***         -0.0209**         -0.0207**         -0.0144***         -0.0136**         -0.0139           (0.00283)         (0.00280)         (0.00844)         (0.00846)         (0.00540)         (0.00546)         (0.00997)           0.0021         0.0022         0.0109*         0.0020         0.00195         0.00222         -0.00150           (0.0022)         (0.00562)         (0.0058)         (0.00452)         (0.00457)         (0.00719)           0.0214***         0.00718         -0.000201         (0.00324)         (0.00324)         (0.00324)           0.119***         0.0411         0.0106         (0.00986)         (0.00986)         (0.00896)         (0.00896)         (0.00844)         (0.0844)         (0.0844)         (0.0844)         (0.0844)         (0.0844)         (0.0844)         (0.0947***         (0.0124)         (0.0268)         (0.0268)         (0.0268)         (0.0268)         (0.0268)         (0.0268)         (0.0268)         (0.0268)         (0.0268)         (0.0268)         (0.0268)         (0.00130)         (0.00268)         (0.00130)         (0.00268)         (0.00130)         (0.00268)         (0.00130)

Note: All regressions flexibly control for distance to the border using the vector of controls  $D_{id}$ : distance\*opiumdistrict, distance\*suitability, distance\*opiumdistrict\*suitability. The variable opiumdistrict is not included as it is collinear with the district fixed effects. Standard errors in parentheses. Standard errors are clustered at the grid-cell level to account for intra-grid correlation. \*\*\* p<0.01, \*\* p<0.05, \* p<0.1

Table 5B: RD estimates at the 10 km bandwidth

Dependent variable:	Lite	racy	Primar	Primary school		Health centre		All-weather road access	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	
opiumdistrict*suitability	-0.0123***	-0.0115***	-0.0208*	-0.0204*	-0.0152**	-0.0144**	-0.0498***	-0.0456***	
	(0.00378)	(0.00374)	(0.0108)	(0.0112)	(0.00709)	(0.00730)	(0.0133)	(0.0133)	
suitability	0.0018	0.0018	0.00854	0.0020	0.0035	0.0040	-0.00647	-0.00380	
	(0.0029)	(0.0029)	(0.00703)	(0.0073)	(0.0061)	(0.0062)	(0.009)	(0.00978)	
aggregate agr. suitability		0.0204***		0.0231**		0.00477		0.0315***	
		(0.00293)		(0.00990)		(0.00444)		(0.0104)	
wheat suitability		0.123***		0.0130		0.00838		0.0544	
		(0.0119)		(0.0373)		(0.0142)		(0.0363)	
soil class clay		-0.142***		-0.216		0.0368		-0.360***	
		(0.0395)		(0.137)		(0.0742)		(0.126)	
log distance to Delhi		0.146***		0.286***		0.0490		-0.138*	
		(0.0218)		(0.0704)		(0.0355)		(0.0715)	
log distance to Calcutta		0.0885***		0.490***		0.0941**		-0.0347	
		(0.0262)		(0.0783)		(0.0371)		(0.0906)	
log distance to Ganges		0.00660***		-0.0289***		0.00119		-0.00880*	
		(0.00128)		(0.00414)		(0.00191)		(0.00473)	
Observations	13,489	13,358	13,489	13,358	13,489	13,358	13,489	13,358	
D <sub>id</sub> controls	X	X	X	X	X	X	X	X	
Colonial district fixed effect	X	X	X	X	X	X	X	X	
R-squared	0.306	0.330	0.098	0.107	0.009	0.010	0.077	0.081	
Mean dependent variable	0.397	0.398	0.757	0.756	0.030	0.030	0.733	0.733	

Note: All regressions flexibly control for distance to the border using the vector of controls  $D_{id}$ : distance\*opiumdistrict, distance\*suitability, distance\*opiumdistrict\*suitability. The variable opiumdistrict is not included as it is collinear with the district fixed effects. Standard errors in parentheses. Standard errors are clustered at the grid-cell level to account for intra-grid correlation. \*\*\* p<0.01, \*\* p<0.05, \* p<0.1

The results for all-weather road access are less conclusive. The coefficient is negative throughout, but only significant at the 10 km bandwidth, where it also larger in magnitude. Appendix Table 1 presents the results for a separate measure of infrastructure that could potentially be more prevalent in poppy-cultivating areas as a result of opium production: irrigation. While critics of the opium trade claimed that the system of advance payments led to indebtedness by farmers or the appropriation of the advance by village headmen, its proponents argued that they allowed farmers to make investments required for poppy cultivation – in particular in wells for irrigation (Allen 1852, Richards 1981). Using 2011 census data on the share of agricultural land that is irrigated, and the share that is irrigated by wells, I find no evidence of a positive lasting impact on irrigation. Instead, the coefficients are negative and significant.

It is important to note that these estimates are local average treatment effects which may not be representative for the entire opium-growing region. Given the empirical design, there are two reasons to question the validity of extrapolating from the results. Firstly, as with any RDD, the restriction of the sample to narrow bandwidths around opium sub-agency borders implies that it may not be representative of villages far from the border. Secondly, the opium suitability index only explains part of the variation in actual opium cultivation – a characteristic common to instrumental variables – which again implies that the results should be interpreted as a LATE.

# 6. Mechanisms

A natural starting point for explaining persistent effects on contemporary measures of education and healthcare is to assess colonial policy in these sectors. Using data on public expenditure between 1895 and 1905 from the district gazetteers of Agra and Oudh, I test whether opium cultivation led to shaped government spending patterns. Table 6 provides OLS estimates (columns 1, 3, 5 and 7) for the effect of (i) being an opium district and (ii) the opium share of agricultural acreage on per capita spending on education and health. Given the likely endogeneity of these variables, I instrument for them using the opium suitability index (columns 2, 4, 6 and 8). All 2SLS results show a significant negative effect of opium production on education and health expenditure. Given the small sample size (48 districts) these results are only suggestive. They are however, consistent with the explanation that current differences in the availability of public goods are rooted in opium-era policies.

Further supporting evidence comes from slightly larger datasets at the tehsil-level. Table 7 replicates the previous analysis for different dependent variables: the number of schools per capita and the number of policemen per capita. Again, the results indicate that opium production was detrimental for human capital provision, with a significant negative effect on the number of schools in all specifications. By contrast columns (5) to (8) show a positive and significant effect on the strength of the police force. This result is in line with Deshpande's (2009) claim that the British Government sought to maintain a large security presence in opium-cultivating areas to enforce the monopsony and curtail smuggling.

Table 6: The effect of opium on education and medical expenditure

Dependent var:	Annual	education ex	xpenditure pe	er capita	Annua	al medical ex	penditure per	capita
	OLS	2SLS	OLS	2SLS	OLS	2SLS	OLS	2SLS
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Opiumdistrict	-0.006**	-0.011**			-0.008**	-0.011**		
	(0.003)	(0.004)			(0.003)	(0.005)		
Opium acreage share			-0.041	-0.459**			-0.174	-0.445*
			(0.117)	(0.210)			(0.137)	(0.227)
Agra prov. fixed effect	0.000	-0.002	0.003	-0.001	0.001	0.000	0.004	0.001
	(0.003)	(0.003)	(0.003)	(0.004)	(0.004)	(0.004)	(0.004)	(0.004)
Provincial capital	0.009	0.009	0.010	0.015	0.007	0.007	0.009	0.013
	(0.008)	(0.008)	(0.009)	(0.010)	(0.009)	(0.009)	(0.010)	(0.010)
Constant	0.037***	0.041***	0.031***	0.038***	0.019***	0.021***	0.013***	0.018***
	(0.003)	(0.005)	(0.003)	(0.004)	(0.004)	(0.006)	(0.004)	(0.005)
N	48	48	48	48	48	48	48	48
F (first stage)		32.55		34.56		32.55		34.56
R-squared	0.173		0.056		0.191		0.093	

Note: Dependent variable calculated as the average per capita expenditure 1895-1905. Model includes a province fixed effect for Agra, with Oudh the omitted category. Provincial capital is a dummy for whether the provincial capital is located in the district. Standard errors in parentheses. \*\*\* p<0.01, \*\*\* p<0.05, \* p<0.1

Table 7: The effect of opium on schools and local police presence

Dependent var:	Numb	er of schools	per 100 inha	bitants	Numb	per of police	per 100 inhal	oitants
	OLS	2SLS	OLS	2SLS	OLS	2SLS	OLS	2SLS
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Opiumdistrict	-0.047***	-0.063***			0.063	0.265*		
	(0.005)	(0.010)			(0.081)	(0.147)		
Opium acreage share			-0.904***	-2.599***			8.709**	11.014*
			(0.265)	(0.485)			(3.528)	(5.931)
Agra prov. fixed effect	0.016**	0.009	0.031***	0.013	-0.226**	-0.125	-0.169*	-0.145
	(0.007)	(0.008)	(0.007)	(0.009)	(0.102)	(0.119)	(0.099)	(0.110)
Provincial capital	0.027	0.027	0.038	0.059**	-0.645**	-0.645**	-0.752**	-0.780**
	(0.021)	(0.021)	(0.024)	(0.026)	(0.319)	(0.321)	(0.318)	(0.321)
Constant	0.186***	0.202***	0.155***	0.184***	1.982***	1.779***	1.895***	1.855***
	(0.008)	(0.011)	(0.008)	(0.011)	(0.117)	(0.169)	(0.103)	(0.131)
N	216	216	216	216	216	216	216	216
F (first stage)		68.52		76.51		68.52		76.51
R-squared	0.365		0.173		0.046		0.070	

Note: Dependent variable calculated as the number of schools or policemen in 1905 divided by the population. Model includes a province fixed effect for Agra, with Oudh the omitted category. Provincial capital is a dummy for whether the provincial capital is located in the district. Standard errors in parentheses. \*\*\* p<0.01, \*\* p<0.05, \* p<0.1

Finally, I evaluate whether the same negative relationship between poppy cultivation and present-day outcomes holds in areas that were formerly princely states. If so, one might conclude that the adverse effects of opium stemmed from its use as a narcotic, rather than the institutions and policies specific to British state-run opium production. In the absence of the geographic boundaries imposed by the British, I cannot replicate the RDD analysis. Appendix Table 2 therefore presents reduced form estimates for a sample of 36,578 villages in 40 princely states which had positive opium cultivation. I include princely state fixed effects and test whether opium suitability explains within-state variation in literacy or public good provision. I find no evidence of a negative impact. For all dependent variables the coefficient is positive and insignificant. Unfortunately, these results are not directly comparable to those in Table 5 and are less well-identified. The absence of an effect in areas not under direct British rule is consistent with the explanation presented in this section but is not conclusive.

# 7. Conclusion

The parts of Bihar, Jharkhand, and Uttar Pradesh where opium was produced under British rule, lag behind much of India in terms of income, literacy, and access to public goods. This paper provides evidence to suggest that the state-run extraction of opium rents causally contributed to these regions' current comparative underdevelopment. Using an RDD design that exploits the interaction between geographic suitability for poppy cultivation and administrative boundaries that confined production to specific areas, I show that the opium industry had persistent negative effects on local development. Villages with a higher likelihood of historical opium cultivation have lower literacy, and less access to private schools and healthcare facilities. There is no evidence that opium cultivation gave rise to persistent benefits in terms of access to roads or irrigation infrastructure. Instead, historical administrative data suggest that British officials in poppy-growing districts invested less in education and health, while spending more on a police force that could help to secure the opium monopsony and combat smuggling. Colonial opium production in India might therefore be considered an example of a historical resource curse. Its adverse effects on the wider economy have persisted long after the opium agencies were closed.

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