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# Returns to Education in the Russian Federation: Variation across Regions

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#### Data and Code

Thanks are due to the Higher School of Economics, Moscow for making the Russian Longitudinal Monitoring Study (RLMS) Household data readily available for researchers around the world. The code used for this paper is made freely available for all researchers at https://bitbucket.org/zagamog/edreru/src/master/

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#### **KEYWORDS**

keyword 1, *keyword* 2, keyword 3, keyword 4, keyword 5, keyword 6, keyword 7

# 1 | REGIONAL RETURNS TO EDUCATION IN THE RUSSIAN FEDERATION: ROLE OF ACCESS TO VOCATIONAL AND HIGHER EDUCATION

#### 1.1 | Data

To estimate returns to education in Russian regions, we use the most recent (2018) round of the Statistical Survey of Income and Participation in Social Programs, collected by Rosstat. The primary purpose of the Rosstat survey was to obtain statistical information, reflecting the role of wages, income from self-employment, property income, pensions, and social benefits in ensuring the material well-being of families. The survey contains data on trends in income and poverty variation among households with different socio-economic status. There are also variables on people's participation in social programs, their pension and health insurance, material and social security of low-income families, and the impact of social policy measures on people's well-being. The sample selected for the empirical modeling is identical to the one used for the RLMS analysis: individuals aged 25-64 who are out of school and have positive labor market experience and income.

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#### 1.2 | Methods

The classical Mincerian equation (described in the previous sections) is the main focus in the regional investigation of returns to education in Russia: in this section we look at how these returns vary across regions. Additionally, we explore the determinants of the established variation through a random effects regression analysis. The equations of interest are as follows:

First level:

$$LogWage_{ij} = b_{0j} + b_{1j} \cdot Educ + b_{2j} \cdot Exp + b_{3j} \cdot Exp^2 + b_{4j} \cdot Gender + \epsilon_{ij}$$

$$\tag{1}$$

Second Level:

$$b_{0j} = \gamma_{00} + \gamma_{0n} \cdot Z + u_{00};$$
  $b_{1j} = \gamma_{10} + \gamma_{1n} \cdot Z + u_{10};$   $b_{ij} = \gamma_{i0} \quad for \quad i \neq 0$  (2)

where an individual i is nested withing a region j, LogWage is a logarithm of monthly wage, Educ stands for highest attained level of education, Exp and  $Exp^2$  reflect the years of working experience and its quadratic term respectively, Gender is a dummy variable for gender, Z is an  $n \times i$  matrix of regional characteristics,  $\epsilon$  and  $u_{00}$ ,  $u_{10}$  are the first-and second-level errors respectively.

The random effects models were estimated using restricted maximum likelihood (REML). Individual Wald tests and likelihood ratio tests were exploited to evaluate the significance of fixed and random effects, respectively. Weights were used in the modeling to ensure the representativeness of the sample across Russian regions (the weighting variable was divided by 1000 to allow the convergence of the multilevel models).

#### 1.2.1 | Left Hand Side (LHS) variable

The outcome to be investigated is the logarithm of monthly monetary remuneration before income tax payment at the main place of work.

#### 1.2.2 | Right Hand Side (RHS) variables

Education, experience, and gender are the first-level variables. Education was included in the random effects models as the focus of interest with a set of regional level variables - equation 2.

The random effects model in our research seeks to estimate the influence of regional level *educational quantity* and *educational quality* measures to explain the variation in education payoffs across Russian regions. To measure educational quantity or access, we used the number of students enrolled in vocational education per 10,000 residents and the number of students enrolled in higher education per 10,000 residents. As a measure of educational quality, standard deviations from the national mean of the Russian school-leaving and university entrance examination, the EGE, were incorporated. We also added variables regarding economic development and the labor market - these are the gross regional product, the level of urbanization, the regional unemployment level and the share of employment in jobs related to natural resources exploitation.

All region-level variables were normalized with Z-standardization before being plugged into the analysis to obtain meaningfully interpretable moderation effects in cross-level interaction models. For the statistically significant interactions, marginal returns to schooling, conditioned on thresholds of region-level characteristics (-1, 0, 1 standard

deviations), were evaluated:

$$\{b_{1j}|Z=1\} = \gamma_{10} + 1 \times \gamma_{1n} \qquad \{b_{1j}|Z=0\} = \gamma_{10} \qquad \{b_{1j}|Z=-1\} = \gamma_{10} - 1 \times \gamma_{1n}$$
 (3)

Table 1.1 demonstrates descriptive statistics of the key variables of interest by regions.

**TABLE 1.1** Descriptive Statistics for Regions in Russia, Rosstat 2018

		Wa	ige	Experie	ence	Edu	cation	, %		der, %
Regions	N	mean	sd	mean	sd	SE	VE	HE	Males	Females
Altayskiy Kray	4646	22127.6	11952.2	23.6	11.0	17.456	54.50	28.05	48.90	51.10
Amurskaya Oblast	2557	33441.2	17409.0	23.2	11.2	16.347	50.65	33.01	49.59	50.41
Arkhangelskaya Oblast	3183	33438.1	16884.2	22.6	10.6	12.692	54.95	32.36	44.17	55.83
Astrakhanskaya Oblast	2836	26474.1	13737.6	23.0	11.3	13.646	55.08	31.28	50.99	49.01
Belgorodskaya Oblast	3692	26281.0	10811.9	23.8	11.1	12.351	54.47	33.18	49.76	50.24
Bryanskaya Oblast	3087	22482.3	9634.1	23.5		19.631				51.34
Chechenskaya Respublika	2010	27718.4	11793.2	18.7	10.6	25.721	26.37	47.91	65.37	34.63
Chelyabinskaya Oblast	6717	27990.8	14280.9	23.9	11.2	12.104	54.53	33.36	47.39	52.61
Chukotskiy Aok	1535	65574.1	32370.8	23.6	10.6	13.941	46.06	40.00	43.97	56.03
Chuvashskaya Respublika	3248	21453.7	12602.2	24.3	11.0	19.119	50.80	30.08	50.18	49.82
Evreyskaya AOb	1536	28532.1	17385.1	23.8	11.2	22.005	50.33	27.67	50.00	50.00
Irkutskaya Oblast	4686	29967.6	17443.1	22.3	11.2	17.520	47.06	35.42	47.57	52.43
Ivanovskaya Oblast	2876	24881.8	12496.8	23.3	10.9	20.341	49.90	29.76	47.77	52.23
Kabardino-Balkarskaya Res.	2006	23592.3	10766.2	21.7	11.6	21.137	40.53	38.33	52.04	47.96
Kaliningradskaya Oblast	2838	29749.2	15489.1	23.5	11.4	13.495	52.40	34.11	50.07	49.93
Kaluzhskaya Oblast	3155	29662.1	12879.5	24.1	11.2	13.312	52.11	34.58	47.92	52.08
Kamchatskaya Kray	2203	51160.5	29997.7	23.1	11.2	13.118	42.99	43.89	47.89	52.11
Karachayevo-Cherkessiya	1510	22900.6	12540.8	22.0	11.8	17.152	40.07	42.78	48.01	51.99
Kemerovskaya Oblast	5056	26287.0	13774.4	23.6	11.3	18.137	52.99	28.88	48.04	51.96
Khabarovskiy Kray	3731	42008.8	21837.8	22.3	11.2	11.900	44.33	43.77	46.15	53.85
Khanty-Mansiyskiy Aok	4335	50837.9	22261.7	22.8	10.5	13.564	46.78	39.65	49.60	50.40
Kirovskaya Oblast	3284	22941.0	13674.6	25.1	11.2	20.128	55.33	24.54	47.69	52.31
Kostromskaya Oblast	2518	23993.1	12090.9	23.6	11.1	12.669	61.28	26.05	47.82	52.18
Krasnodarskiy Kray	8730	32563.7	17499.8	23.0	10.9	15.888	48.57	35.54	50.02	49.98
Krasnoyarskiy Kray	5540	33954.6	21199.2	23.0	11.0	21.588	48.05	30.36	49.64	50.36
Kurganskaya Oblast	2468	20896.9	11539.5	24.4	10.7	21.394	52.47	26.13	48.38	51.62
Kurskaya Oblast	2956	23622.6	11475.0	23.9	11.0	14.783	52.17	33.05	50.30	49.70
Leningradskaya Oblast	4506	32124.3	17227.4	24.2	11.5	7.723	54.77	37.51	46.03	53.97
Lipetskaya Oblast	2869	25037.8	10813.5	24.1	11.0	13.106	53.82	33.08	49.60	50.40
Magadanskaya Oblast	1841	51000.8	23729.4	24.1	11.4	18.523	43.02	38.46	43.24	56.76
Moscow	29921	66263.5	26437.9	20.8	10.8	4.953	32.18	62.86	47.06	52.94
Moskovskaya Oblast	13431	46725.1	20563.7	22.6	11.4	10.975	39.13	49.89	47.51	52.49
Murmanskaya Oblast	3078	43992.5	28841.9	23.4	11.2	12.801	50.45	36.74	49.84	50.16
Nenetskiy Aok	1118	54467.3	23147.1	22.6	10.8	17.263	49.73	33.01	39.98	60.02
Nizhegorodskaya Oblast	6139	30912.9	13291.8	23.4	11.2	16.941	49.31	33.75	47.42	52.58
Novgorodskaya Oblast	2673	26856.0	12683.0	24.6	11.2	15.638	55.74	28.62	45.16	54.84
Novosibirskaya Oblast	5374	29229.9	14687.7	23.9	11.6	16.561	49.33	34.11	47.06	52.94
Omskaya Oblast	3978	25337.5	14613.1	23.6	10.9	22.197	51.31	26.50	51.11	48.89
Orenburgskaya Oblast	4190	24207.0	12519.9	23.3	11.0	15.131	53.68	31.19	51.29	48.71

**TABLE 1.1** Descriptive Statistics for Regions in Russia, Rosstat 2018

		Wa	age	Experi	ence	Edu	cation	, %	Gen	der, %
Regions	N	mean	sd	mean	sd	SE	VE	HE		Females
Orlovskaya Oblast		21901.2		24.7		15.017				53.01
Penzenskaya Oblast		23478.4		24.2		20.722				48.98
Permskiy Krai		29176.6		23.4		13.894				51.83
Primorskiy Kray		37839.9		23.8		14.985				50.02
Pskovskaya Oblast	2382	23838.4	12015.3	25.0		17.632				51.89
Respublika Adygeya	2013	21350.3	10505.9	23.4	11.3	20.666	43.67	35.67	49.53	50.47
Respublika Altay		20285.3		23.0		23.027				56.92
Respublika Bashkortostan	7126	31100.8	15175.2	23.4		12.167				48.02
Respublika Buryatia	2469	29536.3	17237.4	22.1	10.6	17.173	45.61	37.22	48.12	51.88
Respublika Crimea	2895	19916.2	9743.9	22.8		21.244				47.01
Respublika Dagestan	3388	26377.3	11971.9	23.0	10.7	30.519	30.79	38.70	55.99	44.01
Respublika Ingushetiya	1207	23740.2	10168.5	18.2	9.6	10.025	18.89	71.09	61.14	38.86
Respublika Kalmykiya	1751	18568.8	11749.1	23.6	11.4	15.762	40.89	43.35	46.43	53.57
Respublika Karelia	2164	28510.2	16639.5	23.7	10.8	17.144	55.45	27.40	47.00	53.00
Respublika Khakasiya	2064	27288.1	16613.3	23.3	11.1	22.045	51.11	26.84	50.97	49.03
Respublika Komi	2972	35891.6	21554.4	23.8	11.0	16.689	53.47	29.85	46.67	53.33
Respublika Mariy El	2486	21133.1	11941.6	24.1	11.2	18.785	52.98	28.24	47.87	52.13
Respublika Mordovia	2236	21221.0	10837.3	23.1	11.2	15.519	49.11	35.38	48.35	51.65
Respublika Saha (Yakutia)	3243	45763.1	25001.6	23.2	11.3	18.440	45.76	35.80	46.69	53.31
Respublika Severnaya Osetiya	2114	22993.1	12762.5	21.8	11.3	12.677	40.92	46.40	48.91	51.09
Respublika Tatarstan	7212	30327.9	12928.8	23.5	11.1	18.691	48.64	32.67	51.48	48.52
Respublika Tyva	1704	23421.9	16851.3	21.4	10.0	19.777	44.78	35.45	40.43	59.57
Rostovskaya Oblast	6985	28287.2	12779.9	23.1	11.0	15.476	48.03	36.49	50.68	49.32
Ryazanskaya Oblast	2609	25889.2	11760.9	24.7	11.1	12.457	59.37	28.17	49.18	50.82
Saint-Petersburg	11352	48520.8	23771.0	22.8	11.4	5.259	38.15	56.59	46.04	53.96
Sakhalinskaya Oblast	2258	50325.1	25563.0	23.6	11.2	17.493	48.23	34.28	46.94	53.06
Samarskaya Oblast	6275	32584.4	15015.6	23.8	11.1	11.331	47.87	40.80	47.71	52.29
Saratovskaya Oblast	4572	23698.6	12322.4	23.7	10.8	14.961	50.22	34.82	50.42	49.58
Sevastopol	1489	24811.3	13498.9	22.4	11.2	9.671	44.93	45.40	53.32	46.68
Smolenskaya Oblast	2726	25517.8	12104.9	24.6	11.3	14.380	52.31	33.31	46.04	53.96
Stavropolskiy Kray	4945	25263.6	12696.7	22.6	11.3	16.946	43.80	39.25	47.48	52.52
Sverdlovskaya Oblast	7712	35983.2	15242.7	23.6	11.3	16.779	54.94	28.28	48.59	51.41
Tambovskaya Oblast	2781	22698.6	10440.1	24.1	11.0	16.397	53.54	30.06	50.67	49.33
Tomskaya Oblast	3074	29580.6	16745.7	22.1	11.1	13.500	47.56	38.94	46.78	53.22
Tulskaya Oblast	3516	27687.4	11814.7	24.3	11.3	17.491	54.69	27.82	48.98	51.02
Tverskaya Oblast	3157	26310.0	15025.1	25.5	11.1	14.824	56.57	28.60	44.73	55.27
Tyumenskaya Oblast	3095	31441.2	17278.6	22.7	11.2	16.123	52.89	30.99	50.05	49.95
Udmurtskaya Respublika	4073	24044.6	11540.9	23.9	11.3	20.108	51.04	28.85	46.99	53.01
Ul'yanovskaya Oblast	3109	23215.3	10596.4	24.8	10.9	19.170	53.84	26.99	50.37	49.63
Vladimirskaya Oblast	3502	25001.4	12605.8	24.5	11.4	19.503	50.77	29.73	46.49	53.51
Volgogradskaya Oblast	4836	24459.0	12915.8	23.2	11.0	15.881	50.91	33.21	49.69	50.31
Vologodskaya Oblast		28248.9		23.9		17.302				50.39
Voronezhskaya Oblast		26261.9		23.6		22.700				51.63
Yamalo-Nenetskiy Aok		69356.7		21.0		10.683				51.26
Yaroslavskaya Oblast	3361	30261.4	14682.8	24.1		16.215				52.99
Zabaykalskiy Kray		28336.6		23.0		24.561				52.93

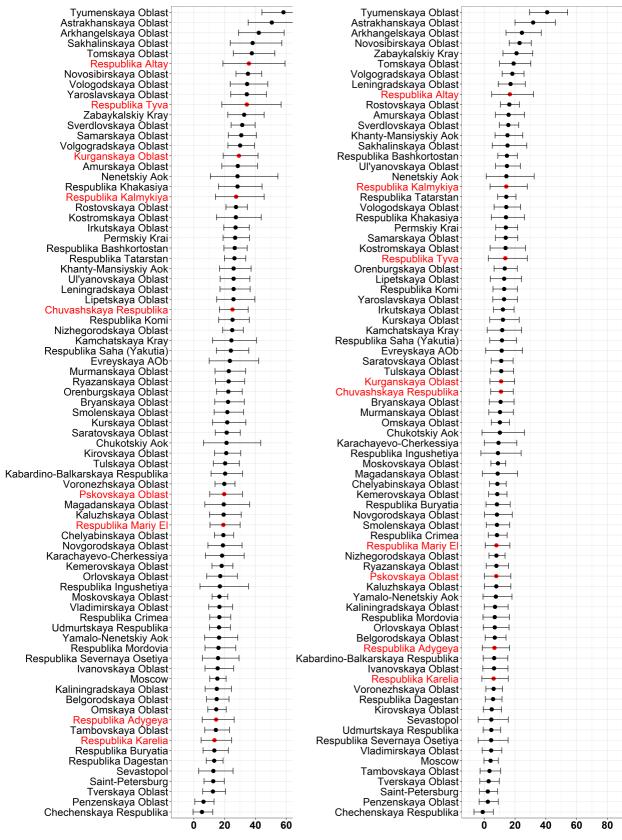
#### 1.3 | Estimation Results of Regional Analysis

A two-level model without covariates was initially fitted and indicated that the Intra-class Correlation (ICC) was equal to 17%, i.e., 17% of the variation in wage outcome was between regions. This is a high enough level to justify the estimation of a random effects model with covariates. Nested models comparison showed that there is a statistically significant regional variation in the effect of education on people's earnings ( $-2 \triangle LL1 = 413.54, p < .001$ ).

We then explored regional diversity in education payoffs in Russia by extracting region-specific estimates of rates of returns to schooling and their respective 95% confidence intervals, using a random slope model (see Figure 1.1). Red color on the picture highlights region called by the Russian Federal Government as priority regions that were ranked lowest regarding poverty, income, and investment climate. A visual inspection of this graph illustrates that premiums to education in Russian regions are rather heterogeneous, varying from 5.2% (Chechenskaya Respublika) to 58.2% (Tyumenskaya Oblast) for university level and from -1% (Chechenskaya Respublika) to 29.6% (Tyumenskaya Oblast) for vocational level.

Next, the possible explanatory factors of the established regional diversity in returns was examined by adding second-level independent variables and their interactions with education levels. The investigation revealed that none of the second-level characteristics are capable of changing the association between education and the amount of money Russian people earn, except for *the coverage by vocational education*. Substantively, it was found that growth in the number of students covered by vocational programs leads to higher schooling premiums concerning both vocational and university education.

Particularly, sufficiently high vocational degree coverage (when this variable is equal to 1 standard deviation) corresponds to the average return rate of 35.8%. Medium vocational degree coverage (when this variable is equal to 0 standard deviation) corresponds to the average return rate of 30.6%. Low vocational degree coverage (when this variable is equal to -1 standard deviation) corresponds to the average return rate of 25.5%. The interpretation of such a finding can imply that this quantity-related dimension of vocational education has the potential to serve as an instrument of boosting financial payoffs from post-secondary education in Russian regions.



### **Appendix**

TABLE A1

	Null model	Mincerian	Random Slope	Cross-Level Interaction
	(1)	(2)	(3)	(4)
Constant	10.178***	10.032***	10.056***	10.065***
	(0.034)	(0.034)	(0.036)	(0.036)
Vocational		0.283***	0.279***	0.267***
		(0.009)	(0.021)	(0.021)
Higher		0.638***	0.641***	0.622***
		(0.009)	(0.025)	(0.025)
Coverage VE X Vocational				0.050**
				(0.025)
Coverage VE X Higher				0.083***
				(0.030)
Experience		-0.026***	-0.027***	-0.027***
		(0.002)	(0.002)	(0.002)
Experience squared		-0.065***	-0.065***	-0.065***
		(0.002)	(0.002)	(0.002)
Females		-0.403***	-0.404***	-0.404***
		(0.005)	(0.005)	(0.005)
Coverage VE			-0.101***	-0.142***
			(0.039)	(0.043)
Variance of Intecept	0.09	0.08	0.09	0.09
Variance of Vocational			0.02	0.02
Variance of Higher			0.04	0.04
Residual Deviance	0.45	0.35	0.34	0.34
sigma	0.67	0.587	0.584	0.584
deviance	119505.212	106528.235	106137.315	106129.127
df.residual	49184	49179	49173	49171
Observations	49,187	49,187	49,187	49,187
Log Likelihood	-59,755.060	-53,289.500	-53,094.620	-53,096.640
Akaike Inf. Crit.	119,516.100	106,595.000	106,217.200	106,225.300
Bayesian Inf. Crit.	119,542.500	106,665.400	106,340.500	106,366.100

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01