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Supplementary Materials for

AI technologies and employee pay in the UK: evidence from matched employer-employee data

The following pages include output from the descriptive comparisons and multilevel regression analyses to supplement the main findings reported in the article. We begin with providing reassurance that imputed data at level two has not distorted the distribution of technology (see table S1 below). Indeed, mean level values for overall digital technology investment and AI equipment are only marginally lower in the imputed data set. Median values and standard deviation for the imputed data also closely resemble those for the non-imputed data.

Table S1: Comparison of Employer AI adoption for imputed and non-imputed data

	Non-imputed	Imputed
Employer Digital technology investment		
mean	0.36	0.34
Std. dev.	0.30	0.30
median	0.29	0.29
Employer adoption of AI equipment		
Mean	0.08	0.06
Std. dev.	0.16	0.14
Median	0.00	0.00
Employer adoption of AI applications		
Mean	0.09	0.09
Std. dev.	0.18	0.17
Median	0.00	0.00

Considering the hierarchical nature of our merged dataset, we used multilevel regression models that partition variance into level one (variance between employees) and level two (variance between industry-region-size clusters) to analyse the relationship between the adoption and employee use of AI technologies and pay. We ran conventional tests to determine whether multilevel (hierarchical) modelling was appropriate. Comparisons between the fixed effects and multilevel maximum likelihood models are reported in Table S2 below. First, the interclass correlation coefficient (ICC1) suggested that around 19 per cent of the overall variation in pay could be attributed to differences across industry-region-firm size clusters. Second, a random intercept multilevel model fared better than a single level model, based on model comparison indices: AIC and BIC.

Table S2: Comparison of multilevel random intercept and single level regression model performance

Outcome variable	Model	AIC	BIC	ICC
Pay	MLM	17357	17376	0.19
	Single level	17863	17877	

The baseline regression models reported in this study are random intercept models with a maximum likelihood estimator. We tested for the applicability of random slope models, but these did not show improvements relative to random intercept regressions. Having fitted the baseline random intercept models with both level one and level two AI adoption and ICT variables we estimated regressions with interaction effects between technology, qualification and employee voice. We further tested for multicollinearity between age and tenure. Correlation between age and tenure is only moderate ($r=0.34$). Ran regression for direct effect and checked for multicollinearity using VIF scores. No score higher than 1.32. This indicates an absence of multicollinearity.

Lastly, we checked for relationship between qualifications and SOC occupational skill groups. Correlation between these two variables is moderate at $r=0.31$. Moreover, treating them as factors and performing cross tabulations reiterates this (see table S3 below).

Table S3: Cross-tabulation occupational skill groups and qualifications

	SOC Skill 1	SOC Skill 2	SOC Skill 3	SOC Skill 4	Total
No qualification	29 (21.0%)	65 (47.1%)	34 (24.6%)	9 (6.5%)	138 (100%)
Other qualification	40 (11.6%)	158 (45.7%)	101 (29.2%)	38 (11.0%)	346 (100%)
Other qualification below A levels or vocational level 3 or equivalent	99 (9.9%)	523 (52.4%)	214 (21.4%)	121 (12.1%)	999 (100%)
A levels of vocational level 3 or equivalent, and above	91 (6.6%)	665 (48.4%)	303 (22.1%)	246 (17.9%)	1373 (100%)
Degree or equivalent, and above	71 (2.7%)	715 (27.5%)	458 (17.6%)	1070 (41.1%)	2604 (100%)

Tables S4 below reports regression estimates for the relationship between technology adoption at levels one and two showing that the measurements derived from employee and employer surveys are indeed correlated with one another.

Table S4: Regression result for the relationship between AI Level 1 and Level 2

<i>Predictors</i>	AI Software Lvl1		AI Hardware Lvl1	
	<i>Estimates</i>	<i>p</i>	<i>Estimates</i>	<i>p</i>
(Intercept)	2.14 (2.08 – 2.20)	<0.001	2.24 (2.18 – 2.30)	<0.001
AI Applications Lvl2	0.37 (0.08 – 0.66)	0.013		
AI Equipment Lvl2			0.58 (0.19 – 0.96)	0.004
Random Effects				
σ^2	1.50		1.63	
τ_{00}	0.19	SectorRegionSize	0.20	SectorRegionSize
ICC	0.11		0.11	
N	581	SectorRegionSize	581	SectorRegionSize
Observations	5343		5354	
Marginal R ² / Conditional R ²	0.003 / 0.116		0.004 / 0.113	

Table S5 reports regression estimates for the relationship between AI adoption and pay.

Table S6 and S7 contain estimates for the interaction effect between AI adoption, occupation and qualification skill groups corresponding to our machine learning analysis in the main article. Table S8 adds the interaction effect with employee voice.

Lastly, Figures S1 and S2 visualise the interaction effects with occupation/qualification groups and employee voice respectively.

Table S5: Regression results for the effect of AI adoption on employee pay

<i>Predictors</i>	pay_numeric			pay_numeric			pay_numeric			pay_numeric		
	<i>Estimates</i>	<i>CI</i>	<i>p</i>	<i>Estimates</i>	<i>CI</i>	<i>p</i>	<i>Estimates</i>	<i>CI</i>	<i>p</i>	<i>Estimates</i>	<i>CI</i>	<i>p</i>
(Intercept)	2.55	2.46 – 2.64	<0.001	-3.07	-3.47 – -2.67	<0.001	2.96	2.89 – 3.04	<0.001	-2.59	-2.98 – -2.20	<0.001
Employee adoption of AI	0.21	0.18 – 0.24	<0.001	0.17	0.14 – 0.20	<0.001						
Age				0.07	0.05 – 0.09	<0.001				0.06	0.05 – 0.08	<0.001
Age square				-0.00	-0.00 – -0.00	<0.001				-0.00	-0.00 – -0.00	<0.001
Hours weekly				0.33	0.28 – 0.38	<0.001				0.35	0.30 – 0.40	<0.001
Skill SOC				0.50	0.46 – 0.54	<0.001				0.52	0.49 – 0.56	<0.001
qualifications numeric				0.12	0.09 – 0.15	<0.001				0.12	0.09 – 0.15	<0.001
size numeric				0.13	0.09 – 0.18	<0.001				0.14	0.10 – 0.19	<0.001
tenure numeric				0.17	0.12 – 0.21	<0.001				0.18	0.13 – 0.23	<0.001
Rep TU numeric				-0.02	-0.09 – 0.05	0.532				0.00	-0.07 – 0.07	0.975
gender2				0.33	0.26 – 0.39	<0.001				0.36	0.30 – 0.43	<0.001
ethniccat4 [Black, Black British, Caribbean or African]				-0.11	-0.31 – 0.09	0.268				-0.06	-0.26 – 0.13	0.520
ethniccat4 [Mixed, multiple or other ethnic group]				0.18	-0.03 – 0.39	0.100				0.23	0.02 – 0.45	0.031
ethniccat4 [White]				0.16	0.03 – 0.29	0.015				0.13	-0.00 – 0.26	0.056
Rep workcouncils numeric				0.26	0.16 – 0.36	<0.001				0.38	0.29 – 0.48	<0.001

Employer adoption of AI	0.66	0.18 – 1.15	0.008	0.33	0.00 – 0.66	0.050
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Random Effects

σ^2	1.47	1.04	1.50	1.05
τ_{00}	0.28 SectorRegionSize	0.08 SectorRegionSize	0.34 SectorRegionSize	0.09 SectorRegionSize
ICC	0.16	0.07	0.18	0.08
N	574 SectorRegionSize	566 SectorRegionSize	576 SectorRegionSize	568 SectorRegionSize
Observations	5036	4538	5184	4676
Marginal R ² / Conditional R ²	0.034 / 0.187	0.381 / 0.424	0.004 / 0.187	0.359 / 0.412

Table S6: Regression results for the moderating effect of qualifications for the AI adoption-employee pay relationship

<i>Predictors</i>	<i>pay_numeric</i>			<i>pay_numeric</i>		
	<i>Estimates</i>	<i>CI</i>	<i>p</i>	<i>Estimates</i>	<i>CI</i>	<i>p</i>
(Intercept)	-2.56	-2.92 – -2.21	<0.001	-1.62	-1.91 – -1.33	<0.001
Employee adoption of AI	0.42	0.32 – 0.52	<0.001			
qualifications numeric	0.25	0.19 – 0.31	<0.001	0.12	0.08 – 0.15	<0.001
Age	0.01	0.01 – 0.02	<0.001	0.01	0.01 – 0.01	<0.001
Hours weekly	0.34	0.29 – 0.39	<0.001	0.37	0.32 – 0.42	<0.001
Skill SOC	0.50	0.47 – 0.54	<0.001	0.53	0.50 – 0.57	<0.001
size numeric	0.14	0.10 – 0.19	<0.001	0.16	0.11 – 0.20	<0.001
tenure numeric	0.20	0.15 – 0.25	<0.001	0.21	0.16 – 0.26	<0.001
gender2	0.34	0.28 – 0.41	<0.001	0.38	0.31 – 0.44	<0.001
ethniccat4 [Black, Black British, Caribbean or African]	-0.09	-0.29 – 0.11	0.371	-0.04	-0.24 – 0.16	0.680
ethniccat4 [Mixed, multiple or other ethnic group]	0.16	-0.05 – 0.38	0.131	0.23	0.02 – 0.45	0.033
ethniccat4 [White]	0.16	0.03 – 0.29	0.019	0.12	-0.01 – 0.26	0.064
Employee adoption of AI × qualifications numeric	-0.06	-0.08 – -0.03	<0.001			
Employer adoption of AI				0.59	-0.40 – 1.58	0.241
Employer adoption of AI × qualifications numeric				-0.05	-0.27 – 0.18	0.667
Random Effects						
σ^2	1.05			1.07		
τ_{00}	0.08	SectorRegionSize		0.10	SectorRegionSize	
ICC	0.07			0.08		
N	566	SectorRegionSize		568	SectorRegionSize	
Observations	4538			4676		
Marginal R ² / Conditional R ²	0.372 / 0.418			0.345 / 0.400		

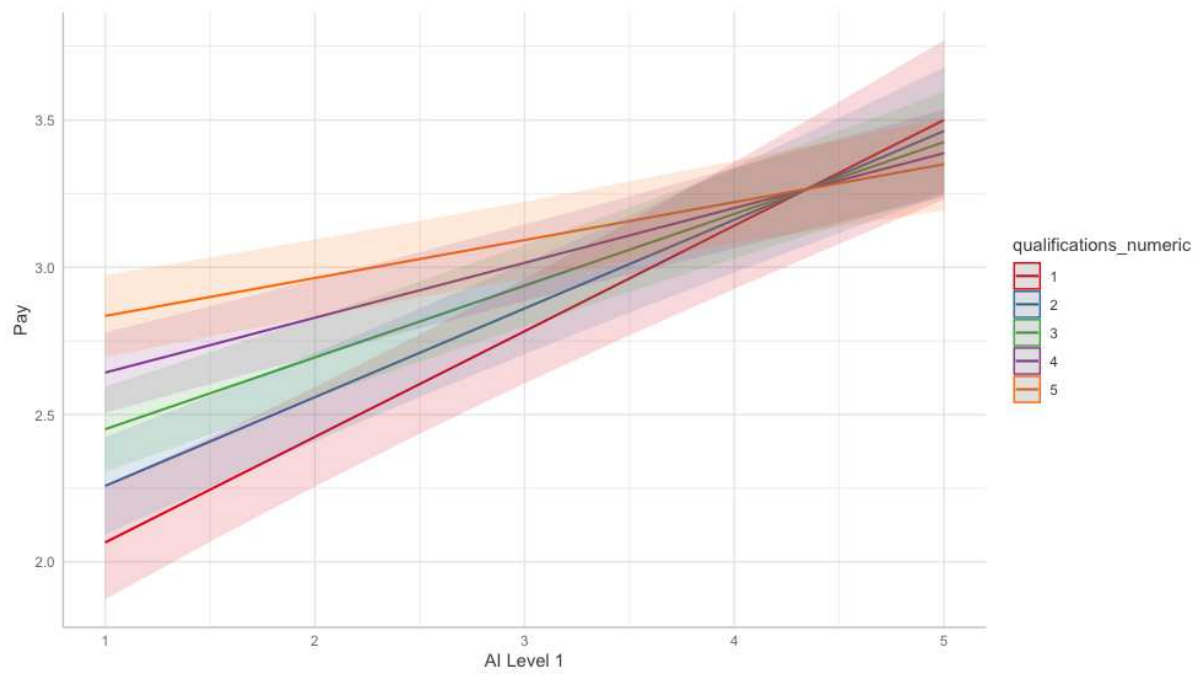
Table S7: Regression results for the moderating effect of occupational skill groups for the AI adoption-employee pay relationship

<i>Predictors</i>	pay_numeric			pay_numeric		
	<i>Estimates</i>	<i>CI</i>	<i>p</i>	<i>Estimates</i>	<i>CI</i>	<i>p</i>
(Intercept)	-2.34	-2.67 – -2.01	<0.001	-1.57	-1.86 – -1.28	<0.001
Employee adoption of AI	0.31	0.23 – 0.39	<0.001			
Skill SOC	0.61	0.54 – 0.68	<0.001	0.52	0.48 – 0.56	<0.001
Age	0.01	0.01 – 0.02	<0.001	0.01	0.01 – 0.01	<0.001
Hours weekly	0.34	0.29 – 0.39	<0.001	0.37	0.32 – 0.42	<0.001
qualifications numeric	0.12	0.08 – 0.15	<0.001	0.11	0.08 – 0.14	<0.001
size numeric	0.14	0.09 – 0.19	<0.001	0.15	0.11 – 0.20	<0.001
tenure numeric	0.20	0.15 – 0.25	<0.001	0.21	0.17 – 0.26	<0.001
gender2	0.34	0.27 – 0.40	<0.001	0.38	0.31 – 0.44	<0.001
ethniccat4 [Black, Black British, Caribbean or African]	-0.09	-0.29 – 0.11	0.388	-0.04	-0.24 – 0.16	0.676
ethniccat4 [Mixed, multiple or other ethnic group]	0.18	-0.04 – 0.39	0.104	0.23	0.02 – 0.45	0.032
ethniccat4 [White]	0.17	0.04 – 0.30	0.012	0.12	-0.01 – 0.26	0.063
Employee adoption of AI × Skill SOC	-0.05	-0.07 – -0.02	0.001			
Employer adoption of AI				-0.18	-1.06 – 0.69	0.677
Employer adoption of AI × Skill SOC				0.19	-0.08 – 0.47	0.164
Random Effects						
σ^2	1.05			1.08		
τ_{00}	0.08	SectorRegionSize		0.10	SectorRegionSize	
ICC	0.07			0.08		
N	566	SectorRegionSize		568	SectorRegionSize	
Observations	4538			4676		
Marginal R ² / Conditional R ²	0.371 / 0.416			0.346 / 0.400		

Table S8: Regression results for the moderating effect of employee voice for the AI adoption-employee pay relationship

<i>Predictors</i>	pay_numeric			pay_numeric		
	<i>Estimates</i>	<i>CI</i>	<i>p</i>	<i>Estimates</i>	<i>CI</i>	<i>p</i>
(Intercept)	-2.55	-3.15 – -1.95	<0.001	-1.61	-2.07 – -1.15	<0.001
Employee adoption of AI	0.41	0.22 – 0.60	<0.001			
voice	0.22	-0.02 – 0.46	0.078	-0.02	-0.18 – 0.13	0.766
Age	0.01	0.01 – 0.02	<0.001	0.01	0.00 – 0.01	<0.001
Hours weekly	0.38	0.32 – 0.44	<0.001	0.40	0.34 – 0.46	<0.001
qualifications numeric	0.11	0.08 – 0.15	<0.001	0.11	0.07 – 0.14	<0.001
Skill SOC	0.50	0.46 – 0.54	<0.001	0.52	0.48 – 0.57	<0.001
size numeric	0.13	0.08 – 0.18	<0.001	0.14	0.09 – 0.19	<0.001
tenure numeric	0.21	0.16 – 0.27	<0.001	0.23	0.17 – 0.29	<0.001
gender2	0.33	0.25 – 0.40	<0.001	0.37	0.29 – 0.44	<0.001
ethniccat4 [Black, Black British, Caribbean or African]	-0.16	-0.40 – 0.08	0.200	-0.12	-0.36 – 0.13	0.356
ethniccat4 [Mixed, multiple or other ethnic group]	0.16	-0.10 – 0.42	0.234	0.24	-0.02 – 0.50	0.074
ethniccat4 [White]	0.15	-0.01 – 0.31	0.063	0.12	-0.04 – 0.27	0.154
Employee adoption of AI × voice	-0.11	-0.20 – -0.02	0.015			
Employer adoption of AI				1.60	-0.61 – 3.80	0.157
Employer adoption of AI × voice				-0.55	-1.49 – 0.39	0.252
Random Effects						
σ^2	1.07			1.10		
τ_{00}	0.07	SectorRegionSize		0.08	SectorRegionSize	
ICC	0.06			0.07		
N	463	SectorRegionSize		465	SectorRegionSize	
Observations	3437			3538		
Marginal R ² / Conditional R ²	0.364 / 0.403			0.339 / 0.385		

Figures S1: Interaction effect between AI adoption and qualification skill groups



Figures S2: Interaction effect between AI adoption and employee voice

