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1 **High-voltage overhead transmission lines and farmland value: evidences from the real estate**
2 **market in Apulia, southern Italy**

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9

10 **Abstract**

11 The construction of high-voltage overhead transmission lines on farmland implies a permanent
12 easement, involving expropriation of land and depreciation of the remaining farm area. The Italian
13 system operator should pay compensation for both aspects, but recognizes only the former. Therefore,
14 landowners often appeal to the law courts and claim compensation for the depreciation of the entire
15 non-occupied area, often obliging the system operator to pay substantial amounts. This delays the
16 provision of new power lines and increases their respective costs.

17 In order to verify the correctness of the *modus operandi* by the system operator and landowners, a
18 study was carried out into the impact of several characteristics of power lines on farmland value in
19 northern Apulia, south Italy. The results highlighted that the area occupied by plinths and cabins, the
20 height of towers and the type of intersection were the main sources of depreciation, which varies
21 depending on the crop. Moreover, depreciation on the residual area exists, but only involves two
22 narrow strips of land on either side of infrastructure. Finally, it is not constant, but tends to decrease
23 rapidly as the distance from the infrastructure increases, and zeroes at 30-70 m from the power line.
24

25 **Keywords:** high-voltage overhead transmission lines; farmland value; easement; compensation;
26 Apulia.
27

28 **1. Introduction**

29 The Italian national electricity grid has a total of 71,000 km of lines, consisting of 49,000 km of
30 150 kV-lines and 22,000 km of 220-380 kV-lines (High-Voltage Overhead Transmission Lines -
31 HVOTLs), with a density of 0.23 km per km² (Terna, 2015). Terna S.p.a. is the national transmission
32 system operator, which owns and manages the power lines, together with their planning and
33 development. Between 1992 and 2015, 220-kV lines decreased by 23%, but 150- and 380-kV lines
34 increased by 15% and 20%, respectively. In Apulia Region, southern Italy, the energy demand has
35 increased by 18% over the last twenty years (Regione Puglia, 2014), thus requiring modernization
36 and improvements to the regional HVOTL grid, including new infrastructures, especially on extra
37 urban areas.

38 The construction of HVOTLs on cultivated land implies several problems: occupation of land by
39 the plinths of towers, electric cabins, etc.; disturbance of agricultural activities for inspection and
40 maintenance; visual and landscape impacts; damage to crops during installation phase; reduced farm
41 profitability; influence on future land uses; depreciation of farmland. Therefore, the assessment of
42 suitable compensations related to these issues is necessary and usually refers to two aspects, namely
43 land expropriation for occupation by power line components and depreciation of the residual (i.e.
44 unoccupied) farmland. The Italian transmission system operator considers only the former, which is
45 calculated as the sum of the following economic items: 100% of the market value of the occupied
46 areas; 25% of the market value of the area of cable projected onto the ground; 7% of the market value
47 of the respect areas. Additional returns are recognized in the case of direct management of farm by
48 the landowner and of voluntary acceptance of the compensation calculated.

49 However, the construction of HVOTLs often also causes depreciation of the remaining farm area.
50 Hence, the following questions:

- 51 a) does the depreciation concern the entire residual area, or only a part of it?
52 b) does the effect of depreciation remain constant over the affected area, or does it progressively
53 decrease to zero at a certain distance?

54 Answering these questions is crucial for fair compensation mechanisms. For example, the impact of
55 power lines on the value of residential properties tends to decrease rapidly as the distance from the
56 power lines increases (Colwell and Foley, 1979; Hamilton and Schwann, 1995), so that only a part of
57 the total area is affected and should be compensated. Nevertheless, as said, the Italian system operator
58 does not include this further effect in assessment of total compensation, so that landowners often
59 appeal to law courts claiming that the entire residual farm area has decreased in value, entitling them
60 to the maximum level of compensation. Finally, the system operator uses subjective criteria to assess
61 the compensation related to land expropriation and does not consider the depreciation of the
62 remaining farmland; landowners, however, subjectively claim the highest compensation on the
63 residual area, forcing the system operator to make substantial payments. This clash of interests causes
64 difficulties in siting (Vajjhala and Fischbeck, 2007), delays in the provision of new power lines, and
65 considerably increases the cost of these infrastructures (Buijs et al., 2011; Cotton and Devine-Wright,
66 2013; Jay, 2004).

67 Only the real estate market of farmlands can provide useful information capable of resolving this
68 conflict, hence the need for accurate studies able to measure the influence of power lines on the value
69 of these properties. Therefore this paper: a) investigated the influence of HVOTL characteristics on
70 farmland value; b) assessed the coefficients related to the impacts of power line characteristics on
71 farmland value, in order to apply them to quantify total compensation for the construction of new
72 HVOTLs; c) studied the extent and trend of depreciation on the residual farmland. In other terms, the
73 paper detects real-estate market-based criteria for the assessment of proper compensations in the
74 presence of HVOTLs on farmland, and at the same time checks the consistency of the economic
75 behaviours by the system operator and landowners. The analysis focused on northern Apulia
76 (southern Italy), and referred to the region's three most important crops (durum wheat, olives and
77 vines).

78 The paper contributes to the literature in two ways. Firstly, no applied economic study has
79 investigated the impact of HVOTL characteristics on farmland value in Italy, or focused on different
80 crops. Secondly, this research helps stakeholders and decision makers with suitable compensation
81 criteria for the construction of new power lines on private farms. Moreover, the findings allow the
82 definition of useful indicators for grid development plans and related strategic environmental
83 assessment.

84

85 **2. Literature review**

86 Electricity transmission via HVOTLs can produce many negative externalities. They include:
87 landscape and visual impacts (Tempesta et al. 2014; Devine-Wright and Batel, 2013); health risks
88 associated with electrical and magnetic fields (Bickel and Friedrich, 2005); damage to the
89 environment and wildlife (Sumper et al., 2010); damage to scientific, historical and cultural areas;
90 reduced profitability of productive activities; land use conflicts (Doukas et al., 2011); decreased
91 property values (Furby et al., 1988). The academic literature about the land depreciation is generally
92 dated and mainly focused on residential areas (Chalmers and Voorvaart, 2009). In particular, when
93 negative impacts are evident, studies have highlighted a discount between 1% and 10% of property
94 value (Pitts and Jackson, 2007; Des Rosiers, 2002), which decreased rapidly as the distance from the
95 power line increased, and usually disappeared at 60-90 meters from the HVOTL (Colwell, 1990). No
96 significant effect was found on the selling prices of vacant residential land with future potential
97 residential use (Blinder, 1979; Kinnard and Mitchell 1988; Mitchell and Kinnard 1996).

98 The effects on the value of properties in rural areas were also uncertain. Several authors (Brown,
99 1976; Chalmers, 2012) found no effects, even where land use was recreational (for example, with a
100 high level of environmental amenity) and rural-residential. On the contrary, Woods (1981) found
101 some effect in a few percentage points, but perhaps related to the fact that these farms would soon
102 become residential areas. Ball (1989) highlighted a 2% reduction, while Jackson and Pitts (2010)
103 assessed a depreciation of 1%-2.5%, although this was considered too small to be statistically
104 significant and attributable to the presence of the line alone.

105 The above studies, which were mostly carried out in the US, highlighted that the impacts of power
106 lines on farmland value are varied and difficult to measure (Pitts and Jackson, 2007), giving

107 conflicting results. Therefore, it is very difficult to generalise, since these studies involve a wide
108 variety of property type and size, market segment, configuration and location, type of power lines
109 and towers, as well as type and quality of the statistical approaches used (Wyman and Worzala, 2013).
110 Moreover, many of these independent studies were not actually independent, but were financed by
111 power line companies (Kinnard, 1990; Kroll and Priestley, 1992), who obviously have an interest in
112 demonstrating a negligible impact on property value.

113

114 **3. Materials and methods**

115 **3.1 Data and variables**

116 The study focused on farmlands in the Municipalities of Ortanova, Foggia, Lucera, Troia and San
117 Severo, in Apulia Region, southern Italy. Over the last decade, there has been significant expansion
118 of HVOTLs in this area, raising issues concerning the payment of unfair compensation and generating
119 conflictual siting, as well as delays in planning and construction of power lines.

120 The most common crops in the area are durum wheat (46.7% of the utilized agricultural area - UAA),
121 olives for oil production (9.8%), and vine (5.4%) (Istat, 2016). Hence, the analysis referred to these
122 crops, characterized by very different quantities of productive factors involved, i.e. land, labour and
123 capital. In particular, durum wheat is characterized by large properties, and small use of labour and
124 capital. Olive groves are related to medium-size properties and moderate recourse to labour and
125 capital. Finally, vineyards have opposite characteristics compared to durum wheat fields.

126 The study was based on a comparison of properties with and without HVOTLs (Table 1), so that
127 two samples were assembled, per crop. Data, collected from real estate transfer acts (62%) and estate
128 agencies (38%), referred to transactions from 2011 to 2016, corresponding to a relatively stable period
129 in the local land market. Using direct interviews with lawyers, brokers and landowners, several
130 variables were identified as influencing the selling price of properties without power lines: the farm
131 area (coded as *Area* and expressed in hectares), able to exploit economies of scale related to labour
132 and capital and directly correlated to the selling price; the yield (*Yield* - tonne hectare⁻¹) for durum
133 wheat fields and vineyards, which influenced revenues and was directly correlated to the selling price;
134 the age of plant (*Age* - Years) for olive groves, which influenced revenues and was directly correlated
135 to the selling price; the orthogonal distance from the farm centre to the nearest HVOTL on nearby
136 properties (*Proximity* - Metres), considered as an indicator of the influence of power lines on
137 surrounding farms, and assumed to be directly correlated to the selling price; the distance between
138 the urban residence of the landowner and its property (*Distance* - Kilometres), indicating more rapid
139 movement to and from the workplace and assumed to be inversely correlated to the selling price; the
140 location of the farm along highways or provincial roads (*Road* - Yes/No), which was an indicator of
141 a faster and easier transport of commodities to and from market places, and was assumed to be directly
142 correlated to the selling price. Finally, the effect of irrigation was not analysed, as it was always
143 present in the olive groves and vineyards, but never in the durum wheat fields, regardless of HVOTLs.

144 The following additional variables were considered for properties with HVOTLs: the ratio between
145 the area occupied by the infrastructure components (plinths, electric cabins, etc.) and the total farm
146 area (*Occupied area* - Percentage) was an important indicator related to the expropriation of land,
147 with further possible implications for farm layout (e.g. changes to farm roads) and managerial
148 strategies (i.e. increased cultivation costs, variations in agronomic practices, etc.). The increase in this
149 ratio was assumed to be directly related to the depreciation of property. The height of towers (*Height*
150 - Metres) was a characteristic of power lines related to the operators' perception of disturbance during
151 their agricultural activities, and was assumed to be positively correlated to land depreciation. The
152 intersection of infrastructure (*Intersection* - Central/Lateral) with farm area was another element of
153 disturbance generated by HVOTLs. In particular, in the interviews landowners considered a central
154 intersection more invasive than a lateral one due to its negative impact on farm layout and cultivation
155 practices. When processing this variable, we assumed that a power line divides a property into two
156 parcels of the same area if the intersection is perfectly central, but of different sizes if the intersection
157 is lateral, thus causing maximum and minimum depreciation, respectively. Therefore, we used the

158 Shannon-Wiener index¹ (Shannon, 1948) in order to characterize the diversity of the parcel areas
 159 generated by the intersection of the property with the power line. The basic idea was that the more
 160 similar the areas of the parcels, the higher the entropy index H , calculated as:
 161

$$162 \quad H = (p_1, \dots, p_s) = -\sum_{i=1}^s p_i \log_2 p_i \quad (1)$$

163
 164 where s was the number of parcels and p_i was the percentage share of the area of the i^{th} parcel
 165 compared to the total farm area. In this study, we considered only traded properties intersected by
 166 just one infrastructure, excluding farms with two or more power lines (only two survey cases), so that
 167 s was always equal to two and $p_1 + p_2$ was the total farm area. The Shannon-Wiener index assumed
 168 nominal values between zero (absence of entropy), indicating a power line on the boundary, and $\frac{1}{2}$
 169 (maximum entropy) for a central power line dividing the total area into two equal parcels. Therefore,
 170 this variable was assumed to be positively correlated to land depreciation. Moreover, this approach
 171 was considered more accurate than the use of a dichotomous variable, as the attribution of discrete
 172 values (0 or 1) would have been affected by high subjectivity. On the contrary, through the continuous
 173 index used, we objectively measured the level of intersection within properties.
 174 Finally, in order to examine the spatial trend of depreciation across the residual area, we calculated
 175 the orthogonal and greatest distance from the central point of the line to the opposite boundary
 176 (*Boundary - Metres*). In this case, it was assumed that the longer the distance, the smaller the
 177 depreciation.

178 For the variables concerning the intrinsic characteristics of the properties, i.e. *Area*, *Yield*, *Age* and
 179 *Road*, data were collected through the real estate sources already mentioned, while the *Distance* and
 180 the *Proximity* variables were measured using Google Maps, from February to October 2016. The
 181 variables related to the characteristics of the power lines (*Occupied area*, *Height*, *Intersection* and
 182 *Boundary*) were calculated via overlapping between the cadastral maps and the respective Google
 183 Maps in ArcMap 10.4, between November 2016 and April 2017. Any missing data (often concerning
 184 the yield, the age of plants and the height of towers) was supplied by direct inspections on properties.
 185

186 3.2 The model

187 The analysis was based on the hypothesis that the selling price of farmland without HVOTLs was
 188 only influenced by its intrinsic characteristics (*Area*, *Yield* or *Age*, *Proximity*, *Road*, *Distance* and
 189 *Proximity* variables). In the presence of HVOTLs, instead, the selling price was also affected by
 190 additional factors related to power lines (*Occupied area*, *Height*, *Intersection* and *Boundary*
 191 variables).

192 A linear form of the model based on the ordinary least squares (OLS) (Wooldridge, 2012) was
 193 assumed, except for the *Boundary* variable. In this last case, we assumed the maximum depreciation
 194 next to the HVOTL, rapidly decreasing down to zero as the distance from the power line increased,
 195 so that a reciprocal relationship was used (Colwell, 1990). Hence, the following model was
 196 implemented:
 197

$$198 \quad \ln SP_i = \beta_0 + \sum_{j=1}^n \beta_j x_{ij} + \sum_{l=1}^m \beta_l z_{il} + \beta_B (1/Boundary_i) \quad (2)$$

199
 200 where: SP_i was the selling price of the i^{th} property and \ln was its natural log; x_{ij} was the j^{th}
 201 characteristic (*Area*, *Yield* or *Age*, *Proximity*, *Distance*, *Road* and *Proximity* variables) of the i^{th}
 202 property; z_{il} was the l^{th} characteristic (*Occupied area*, *Height* and *Intersection* variables) of the
 203 HVOTL in the i^{th} property; *Boundary* was the distance between the power line and the boundary of

¹ The Shannon-Wiener index, based on the concept of entropy, is a diversity index that reflects how many different types (in this case the farmlands) there are in a dataset, and simultaneously takes into account how evenly the entities (in this case the areas of the parcels generated by the intersection with the power line) are distributed among those types.

204 the i^{th} property; $\beta_0, \beta_j, \beta_l$ and β_B were the unknown coefficients to be estimated.

205 We used the semilog functional form, from which, if the assumptions of the classical linear
206 regression model are fulfilled², the OLS method allows us to estimate the parameters (Gujarati and
207 Porter, 2009). In particular, the OLS estimators $\hat{\beta}$ s obtained will be best linear unbiased estimators
208 of β s. Moreover, the percentage variation in SP for an absolute change in each regressor can be
209 calculated by multiplying the relative change in SP by 100. That is, $\hat{\beta} \times 100$ gives the variation rate
210 in SP . This model allowed the direct assessment of the depreciation rates related to the characteristics
211 of HVOTLs, thus providing practical information for stakeholders and policy makers, in order to
212 evaluate compensation for landowners due to the construction of new power lines.

213 The choice of this functional form was verified by the MacKinnon's J-test. Firstly, the null hypothesis
214 that the hedonic form was linear against the log-linear alternative hypothesis was tested. In a first
215 step, the log-linear model was estimated and the predicted values were calculated. In the second step,
216 the linear model was estimated, with the predicted values from the first regression included as extra
217 regressors. The t-statistic for the predicted values, i.e. the test statistic for the J-test, was 11.36, so that
218 the null hypothesis of a linear functional form was rejected. The analogous test for the log-linear
219 functional form against the linear functional form provided a t-statistic of 1.63, hence the null
220 hypothesis of a log-linear functional form was accepted.

221 A preventive multicollinearity analysis based on the correlation coefficient was also carried out in
222 order to identify possible overlap effects among the independent variables.

223 Finally, in regression analysis, the selection of informative variables able to explain the obtained
224 model is crucial, since too many regressors may lead to non-robust parameter estimates and non-
225 interpretable models. In this study, we performed the OLS-lasso (least absolute shrinkage and
226 selection operator) regression (Efron et al., 2004; Tibshirani, 2011, 1996), i.e. a variable shrinkage
227 based on a penalty approach for improving OLS. The lasso model is interpretable like the subset
228 selection method, and is as stable as ridge regression (Hastie et al., 2009). Moreover, the lasso
229 regression is a tolerant method for reducing the sensitivity of regression parameters to
230 multicollinearity.

231 Generally, in addition to the sum of squares error minimization, the lasso model considers L_1 norm
232 constraint on the sum of the absolute value of the regression coefficients, so that some of these latter
233 shrink to exactly zero and are excluded from the analysis. OLS estimates have low bias but high
234 variance of the predicted response, which decreases following removal of predictors from the full
235 model, although the bias increases as trade-off (Miller, 2002). Therefore, OLS accuracy can be
236 improved by shrinking or setting some coefficients to zero, so that this bias–variance trade-off often
237 allows better predictions. The approach is also useful if some predictors are highly correlated, as the
238 lasso model picks only one of them and shrinks the others to zero.

239 In formal terms, \mathbf{X} denotes the $n \times p$ matrix of predictors, whose elements are centred and scaled
240 to have unit standard deviation and mean zero, and \mathbf{y} denotes the $n \times 1$ response vector, whose
241 elements have mean zero. If $\boldsymbol{\beta}$ is a $p \times 1$ vector of regression coefficients, for a given parameter λ , the
242 lasso regression coefficients are the solution to the constrained optimization problem:
243

$$244 \hat{\boldsymbol{\beta}}_{Lasso} = \underset{\boldsymbol{\beta}}{\operatorname{argmin}} \|\mathbf{y} - \mathbf{X}\boldsymbol{\beta}\|^2 \quad \text{s.t.} \quad \sum_{j=1}^p |\beta_j| \leq \lambda \quad (3)$$

245

² The following assumptions were verified using appropriate tests, selected from those reported in literature also according to the small size of the samples (Gujarati and Porter, 2009):

- 1) Normality of residuals: Shapiro–Wilk test (H_0 : residuals are normally distributed);
- 2) Homoscedasticity of residuals: Breusch-Pagan test (H_0 : residuals have constant variance);
- 3) No autocorrelation of residuals: Durbin–Watson test (H_0 : residuals are not autocorrelated, if $d_U < d < 4 - d_U$).

The assumptions concerning the multicollinearity among predictors and the functional form of the model were tested and commented in the text.

246 where $\lambda \geq 0$ is a tuning parameter controlling the amount of shrinkage, and, provided that it is small
247 enough, some regression coefficients will be exactly zero, thus allowing the automatic selection of
248 the most informative predictors (i.e. the ones with nonzero regression coefficients). In other terms,
249 the parameter λ tunes estimate results through a trade-off between accuracy in prediction and number
250 of predictors, so that lasso selects a subset of regression coefficients for each value of λ .

251 In order to select the proper λ , thus ensuring an accurate and interpretable model, the lasso solutions
252 for different values of λ are tested by model selection optimality criteria. i.e. k -fold cross validation
253 (CV) (Efron et al., 2004), or information criteria, namely AIC or BIC (Zou et al., 2007; Tibshirani
254 and Taylor, 2012). However, in our experience of real estate assessment, also confirmed in other
255 research fields (Mauerer et. al, 2015), BIC and CV identify the most and least sparse model,
256 respectively. AIC, instead, allows a desirable compromise, therefore we preferred this optimality
257 criterion. The appropriate model for each crop was selected on a grid of 100 λ -values, and the best
258 was chosen according to the AIC-criterion. Standard errors were computed via bootstrap based on
259 500 bootstrap samples. In turn, approximate p values based on simple, two-sided t tests were
260 calculated by the bootstrapped standard errors.

261 The variance inflation factor (VIF) was calculated to further verify the presence of
262 multicollinearity among the predictors. Firstly, we ran the lasso regression in which each nonzero
263 regressor in the equation (2) was a function of all the other nonzero covariates. Then, we calculated
264 the VIF factor through the following formula (Wooldridge, 2012):
265

$$266 \quad VIF = \frac{1}{1 - R_i^2} \quad (4)$$

267
268 where R_i^2 was the coefficient of determination of the regression equation calculated in the first step.

269 **4. Results**

270 **4.1 Sample characteristics**

271 The descriptive statistics (Table 1) highlighted that the presence of HVOTLs on farmland influenced
272 selling prices, with depreciations of 6.3% for durum wheat fields, 9.3% for olive groves and 14.4%
273 for vineyards. Moreover, durum wheat fields were much larger than olive groves and vineyards, as
274 expected. Finally, significant differences emerged also between farm surface area, distance from
275 urban residence and HVOTLs, so that the power lines were located far from cities, i.e. in areas where
276 farms are larger, mainly due to a reduced level of land fragmentation.
277

278
279 Table 1 – Descriptive statistics of the samples, per crop.
280

281 **4.2 Lasso results**

282 The multicollinearity analysis indicated correlation coefficients between 0.03 and 0.18,
283 highlighting the absence of a notable overlap among the independent variables, for each crop. The
284 absence of multicollinearity was also confirmed by the VIF values (Table 2), which, for the nonzero
285 predictors, were much lower than the thresholds commonly used by analysts, namely 5 or even 10,
286 according to Snee (1973) and Marquandt (1980), respectively. Regarding the other assumptions, the
287 respective tests indicated normal, homoscedastic and uncorrelated residuals. All the estimated models
288 (Table 2) had adjusted R^2 from 74% to 86%, and the significance of the F-tests was 1%, indicating
289 good general fittings. All selected predictors had high significance (at 1% and 5%), except for the
290 *Road* variable in durum wheat fields with power lines.

291 The lasso estimates pointed out a different influence of regressors on selling prices. In particular,
292 in the absence of HVOTLs, the farmland size positively influenced durum wheat fields and vineyards,
293 due to benefits related to economies of scale, mainly related to mechanization. However, this variable
294 had no effect for the olive groves, probably because of the family management of this crop in the
295 territory considered, and the consequently small size of the farms (under three hectares). The variables
296 related to profitability (yield for durum wheat fields and vineyards, and age of plants for olive groves)

297 had the expected positive effect, while the distance from the urban residence of the landowner
298 negatively influenced the selling price of olive groves and vineyards, but was irrelevant for durum
299 wheat fields. Indeed, the wheat cultivation system involves extensive practices, which imply a low
300 intensity of capital and labour, so that the variable *Distance* did not affect the agronomic practices
301 and then property value. Closeness to principal roads positively influenced the property value for all
302 crops, due to faster and easier transportation of commodities, whereas the proximity of farms to
303 HVOTLs on nearby properties did not affect the selling price. This could be due to the relatively large
304 average distances in the samples, namely 319 m for the durum wheat fields, 214 m for the olive groves
305 and 441 m for the vineyards. Some effect might have emerged with smaller distances, but would
306 probably not have been significant.

307
308 Table 2 – Results of the lasso analysis.

309
310 In the presence of power lines, the significance of the variables related to the intrinsic
311 characteristics of farmlands generally decreased, and some of them often exited from the models. In
312 any case, their impact respected the expected signs. In particular, among the intrinsic characteristics
313 of properties, the selling price for durum wheat fields and olive groves was affected only by the
314 variables related to production, namely *Yield* and *Age*. The *Area* variable remains in the model for
315 vineyards, however, indicating the great importance of economies of scale generated by the predictor
316 for this crop.

317 On the contrary, selling price was mainly influenced by HVOTL characteristics, but always in
318 negative way, thus generating depreciation. The type of intersection was the most influential variable,
319 causing a fall in property values from 10% for the durum wheat fields to 27% for vineyards, per each
320 unit value of the Shannon-Wiener index. The occupied areas by power line components reduced the
321 selling price of 1% for all the considered crops, depreciation corresponding to the market value of
322 land, and the height of towers influenced the value of properties between 0.03% and 0.07% per metre.
323 The negative impact of the *Boundary* variable on selling price was between 6% for durum wheat
324 fields and 15% for vineyards, per each reciprocal value of distance ($1/\text{Boundary}$).

325 Finally, the analysis highlighted the following findings. In general, the presence of power lines
326 sensibly reduced the influence of the agronomic variables (farmland area, yield, age of plants,
327 distance from the urban residence of landowner and proximity to principal roads) on property value,
328 above all in case of extensive crops (Durum wheat). However, their expected signs remain.
329 Furthermore, the area occupied by plinths and cabins, the height of towers and the type of intersection
330 were the main causes of depreciation. Specifically on the first aspect, which generates land
331 expropriation, the depreciation was assessed equal to 1%, i.e. to the market value of land, so that the
332 first part of the criterion used by the system operator is valid. However, the other compensations
333 recognized, namely 25% of the market value of the area of cable projected onto the ground and 7%
334 of the market value of the respect areas, were devoid of economic justification. Other items, instead,
335 should be considered, namely the compensations related to the height of towers and to the type of
336 intersection. Therefore, the total depreciations resulting from both the land expropriation and the
337 characteristics of power lines are depicted in table 3.

338
339 Table 3 – Depreciations per characteristic of HVOTL and crop.

340
341 On the residual farmland, the analysis highlighted the presence of depreciation, which however was
342 not constant over the whole property. Considering the specific functional form, the impact of HVOTL
343 on the unoccupied area zeroed at approximately 30 m for durum wheat fields, 40 m for olive groves
344 and 70 m for vineyards (Figure 1). This tendency was due to the increasing use of capital and labour
345 factors which durum wheat fields, olive groves and vineyards respectively request. In particular, the
346 presence of HVOTL increases the negative perception of risk for winegrowers, who spent over 400
347 hours/hectare/year on the agronomic activities (compared to 200 hours/hectare/year for olive groves
348 and only 20 hours/hectare/year for durum wheat fields). In addition, the presence of HVOTLs on
349 vineyards makes it more difficult to perform mechanized agronomic activities, while the construction

350 of new power lines creates difficulties because it necessitates changes in the structure of these
351 cultivation systems. These issues, however, were less important for olive groves and almost
352 insignificant for durum wheat fields, so that the impact of power lines on these last production systems
353 was smaller and less extensive. In any case, the influence of HVOTLs was confined to each property
354 and, therefore, power lines did not affect the neighbouring farmland (*Proximity* variable).

355 These findings could be used for fair criteria to assess the depreciations for landowners. In this
356 connection, authorities could apply schedule percentage values based on the marginal effects
357 assessed.

358
359 Figure 1 – Depreciation percentage based on the distance from the power line.
360

361 **5. Discussion and policy implication**

362 The study pointed out a considerable impact of HVOTLs on farmland, differentiated per crop.
363 Earlier hedonic studies³ showed that the presence of lines on urban properties negatively affected
364 land value (−1% to −10%). In the present study, the depreciations were sizeable, considering the
365 agricultural use of the properties analysed. These high percentages could be the combined effect of
366 several aspects, namely visual impact, perception of health risk due to electrical and magnetic fields
367 and to noise, in addition to farm management issues. The results concerned power lines already on
368 farmland, but they can also be applied to the construction of new HVOTLs, which leads landowners
369 to claim very high levels of compensation, mostly for depreciation of the residual farmland. This
370 attitude could be explained by several strategic behaviours, including the “Not-In-My-Backyard”
371 (NIMBY) reaction, concerning public opposition to the siting of HVOTLs (Kyle, 2013; Cain and
372 Nelson, 2013). Another more rigorous scientific behaviour is “place attachment”, i.e. the sense of
373 affiliation toward the physical location where people live or work (Joe et al., 2016; Aas et al., 2014;
374 Cotton and Devine-Wright 2013; Devine-Wright, 2013). In this case, opposition to siting HVOTLs
375 is related to demographic factors, i.e. gender, age and education, but above all to specific
376 characteristics of the project, i.e. positive or negative impacts, procedural justice, trust in the
377 developer, etc. These aspects, together with risk perception, noise and farm management issues, could
378 lead landowners to request sizeable compensation for the construction of new HVOTLs on their
379 properties (Acciani and Sardaro, 2014), especially for the residual farm area. However, this analysis,
380 which was based on land market, and therefore on the actual behaviour of operators, has demonstrated
381 that when HVOTLs are already present on farmland, the depreciated area consists of just two narrow
382 strips of land (30-70 m depending on the crop) on either side of the power line. Hence, landowners’
383 claims for the maximum amount of compensation concerning the entire remaining area are
384 unjustified. However, the *modus operandi* of the system operator also appears unfair, as some
385 depreciation on the remaining farmland should be recognized in any case.

386 These findings have highlighted that the approach of landowners differs before and after
387 construction of a HVOTL. This could be explained by the time factor, since several studies (Chalmers
388 and Voorvaart, 2009; Pitts and Jackson, 2007) have reported that the impact of recent power lines
389 built on private properties was initially significant, but quickly diminished over time.

390 The compensatory mechanism in the planning process for power lines should be based on the real
391 estate market, which reflects the true attitudes of landowners. Keeping in mind this assumption, the
392 analysis showed a large degree of inconsistency between their claims during the siting phase and their
393 actual behaviour on the real estate market regarding farmland with pre-existent HVOTLs. On the
394 other hand, depreciation of a part of the remaining area should be recognized by the system operator.
395 The investigation of real-estate characteristics of farmlands with HVOTLs is a crucial element for
396 fair and quicker planning process, therefore these findings should be considered in negotiations
397 concerning land use rights. On the contrary, the absence of real-estate surveys for HVOTL siting can
398 increase the risk of project failure and/or additional costs. These issues could stem from a range of

³ A hedonic study assumes that the good being valued is composed by constituent parts, each of which is valued by market. The contributory value of each characteristic is determined through hedonic models based on regression analysis. Hedonic models are commonly used in real estate economics and Consumer Price Index calculations.

399 factors: lack of trust in the public or private agencies responsible for siting, construction and
400 management of lines (Ceglarz et al., 2017; Jenkin-Smith et al., 2009; Schively, 2007); implementation
401 of non-optimal technological solutions (Lienert et al., 2015); lack of information and notification
402 (Lienert et al., 2018); a sense of exclusion from decisional process by people (Gross, 2007); a negative
403 institutional context (Friedl and Reichl, 2016; Devine-Wright, 2009), and a lack of simplified and
404 standardised regulatory frameworks (Battaglini et al., 2012). In other words, it is necessary to deal
405 with issues related to societal, market and political acceptance (Devine-Wright et al., 2017). By
406 ensuring open participation, as well as collaborative and coordinated planning supported by economic
407 findings, conflicts can significantly reduce (Ritchie et al., 2013), enabling greater trust and better
408 policy outcomes (Steinbach, 2013; Beierle and Konisky, 1999). The transmission system operator
409 should implement well-structured and clear processes involving transparent communication
410 (Ciupuliga and Cuppen, 2013). This approach could improve participants' knowledge and their sense
411 of procedural fairness, benefiting from the effects of "procedural justice" and "social trust" (Schively,
412 2007; Mohanty and Tandon, 2006; Roussopoulos, 2005).

413

414 **6. Conclusions**

415 Real estate studies on the influence of HVOTL characteristics on farmland value can facilitate
416 transmission line planning, construction and management outside urban centres in Italy. This research
417 pointed out three important aspects. The first one concerns the Italian system operator, which should
418 resort to assessment criteria based on scientific findings. Therefore, the compensation to landowners
419 should concern land expropriation, height of towers, type of intersection and depreciation of a part of
420 the unoccupied area. The remaining two aspects concern landowners. During the siting phase, based
421 on individual sentiment, social interaction and political context, they claim sizeable compensations
422 on the total unoccupied area, so to conflict with the system operator. However, this study revealed
423 that, when landowners trade properties with pre-existent power lines, the impact of HVOTL is
424 strongly reduced, above all on the residual area. Hence, landowners claim compensations which are
425 not supported by the real estate mechanisms which themselves trigger. These findings can enhance
426 the regulatory frameworks and improve the energy supply to community by involving public
427 cooperation based on impartial and transparent information.

428 However, real estate analyses in Italy are difficult to carry out, for lack of transparency of real
429 estate market, scarce vivacity of land transactions, and absence of a public agency in charge of the
430 gathering and management of the trade information. Moreover, these analyses also request
431 multidisciplinary data concerning the characteristics of power lines and their location, which however
432 are not made available by institutions. Finally, to date, only direct and lengthy (in this case more than
433 one year) surveys make possible the assessment of fair compensations. On the contrary, the creation
434 of a single agency in charge of the collection and management of real estate and power-line
435 information on the national territory is desirable, thus allowing also a periodical data updating for the
436 increase of the quality of valuations over time.

437

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Table 1 – Descriptive statistics of the samples, per crop.

Variable	Unit	Code	Durum wheat fields								t-stat sign.
			Without HVOTL (n=71)				With HVOTL (n=59)				
			Min.	Max.	Mean	S. d.	Min.	Max.	Mean	S. d.	
Selling price	€ ha ⁻¹	Price	22,910.70	31,031.26	26,832.15	11,820.48	21,346.08	27,721.04	25,192.17	9,032.48	**
Farm area	ha	Area	3.62	126.26	18.23	14.39	4.12	138.91	20.47	11.98	**
Yield	t ha ⁻¹	Yield	2.85	4.12	3.42	1.76	2.78	3.96	3.7	1.53	
Age of plants	Years	Age	-	-	-	-	-	-	-	-	-
Proximity	m	Proximity	222	1,136	318.60	396.00	-	-	-	-	-
Distance from urban residence	km	Distance	5	13	8.80	5.90	6	13	12.06	7.67	**
Highway or provincial roads	Yes/No	Road	0	1	0.40	0.49	0	1	0.38	0.28	
Occupied area/total farm area	%	Occupied area	-	-	-	-	0.84	5.55	4.81	2.43	-
Height of towers	m	Height	-	-	-	-	18.00	50	32.55	9.23	-
Intersection	Shannon-Wiener index	Intersection	-	-	-	-	0.12	0.43	0.22	0.28	-
Distance from boundary	m	Boundary	-	-	-	-	36.04	319.63	125.68	66.63	-
Variable	Unit	Code	Olive groves								t-stat sign.
			Without HVOTL (n=56)				With HVOTL (n=42)				
			Min.	Max.	Mean	S. d.	Min.	Max.	Mean	S. d.	
Selling price	€ ha ⁻¹	Price	26,338.04	45,481.26	36,529.03	13,0372.88	28,193.13	37,718.90	33,160.83	24,779.35	***
Farm area	ha	Area	0.77	9.13	2.94	3.60	1.01	12.36	3.13	2.58	*
Yield	t ha ⁻¹	Yield	-	-	-	-	-	-	-	-	-
Age of plants	Years	Age	50.17	85.10	71.00	41.40	54.13	91.49	68.88	29.03	
Proximity	m	Proximity	108	1,452	214.29	338.18	-	-	-	-	-
Distance from urban residence	km	Distance	2.11	15.6	9.29	3.82	4.91	14.02	12.46	6.91	**
Highway or provincial roads	Yes/No	Road	0	1	0.27	0.17	0	1	0.26	0.20	
Occupied area/total farm area	%	Occupied area	-	-	-	-	1.72	9.81	3.14	3.20	-
Height of towers	m	Height	-	-	-	-	18	50	20.2	18.07	-
Intersection	Shannon-Wiener index	Intersection	-	-	-	-	0.11	0.37	0.26	0.16	-
Distance from boundary	m	Boundary	-	-	-	-	65.39	172.4	100.57	112.09	-
Variable	Unit	Code	Vineyards								t-stat sign.
			Without HVOTL (n=49)				With HVOTL (n=41)				
			Min.	Max.	Mean	S. d.	Min.	Max.	Mean	S. d.	
Selling price	€ ha ⁻¹	Price	39,420.22	57,851.03	45,119.92	18,032.10	31,482.55	52,992.48	38,660.21	26,794.35	***
Farm area	ha	Area	1.86	11.65	4.90	2.43	2.13	10.47	4.12	3.66	*
Yield	t ha ⁻¹	Yield	14.37	36.28	19.51	11.74	12.57	35.18	18.83	12.62	
Age of plants	Years	Age	-	-	-	-	-	-	-	-	-
Proximity	m	Proximity	217	1,361	440.82	476.01	-	-	-	-	-
Distance from urban residence	km	Distance	3	18	14.63	12.95	5	21	16.22	10.05	*
Highway or provincial roads	Yes/No	Road	0	1	0.44	0.31	0	1	0.51	0.25	
Occupied area/total farm area	%	Occupied area	-	-	-	-	0.09	13.25	2.15	3.62	-
Height of towers	m	Height	-	-	-	-	18	50	29.69	12.18	-
Intersection	Shannon-Wiener index	Intersection	-	-	-	-	0.13	0.32	0.19	0.20	-
Distance from boundary	m	Boundary	-	-	-	-	18.04	236.52	113.96	83.17	-

***Sign. 1%; ** Sign. 5%; * Sign. 10%

Table 2 – Results from the lasso analysis.

Variable	Durum wheat fields						Olive groves						Vineyards					
	Without HVOTL			With HVOTL			Without HVOTL			With HVOTL			Without HVOTL			With HVOTL		
	Coeff.	P value	VIF	Coeff.	P value	VIF	Coeff.	P value	VIF	Coeff.	P value	VIF	Coeff.	P value	VIF	Coeff.	P value	VIF
Intercept	10.0867	0.000	0	10.1154	0.000	0	9.9865	0.008	0	10.1898	0.005	0	10.1739	0.000	0	10.3184	0.002	0
Area	0.0014	0.021	1.1332	0			0			0			0.0062	0.003	1.3406	0.0055	0.015	1.1216
Yield	0.0113	0.002	1.0557	0.0090	0.038	1.0602	-	-		-	-		0.0116	0.004	1.1009	0.0114	0.036	1.0332
Age	-	-		-	-		0.0053	0.001	1.0048	0.0052	0.002	1.0068	-	-		-	-	
Distance	0			0			-0.0006	0.028	1.1875	0			-0.0056	0.030	1.2715	0		
Road	0.1275	0.003	1.1095	0.0720	0.085	1.1140	0.0617	0.032	1.1932	0			0.1489	0.017	1.0098	0		
Proximity	0			-	-		0			-	-		0			-	-	
Occup. area	-	-	-	-0.0105	0.004	1.1631	-	-	-	-0.0117	0.000	1.1418	-	-	-	-0.0119	0.000	1.1805
Height	-	-	-	-0.0003	0.033	1.2285	-	-	-	-0.0004	0.016	1.2055	-	-	-	-0.0007	0.002	1.2147
Intersection	-	-	-	-0.1015	0.000	1.0716	-	-	-	-0.1629	0.000	1.1184	-	-	-	-0.2694	0.000	1.0596
Boundary	-	-	-	-0.0578	0.019	1.1322	-	-	-	-0.1020	0.041	1.1705	-	-	-	-0.1468	0.000	1.1742
R ²	0.8415			0.8225			0.7882			0.7719			0.8726			0.8352		
R ² adj	0.8291			0.8054			0.7666			0.7393			0.8574			0.8110		
F-value	22.90	0.000		18.49	0.000		27.88	0.000		18.31	0.000		28.70	0.000		19.90	0.000	
N.	71			59			56			42			49			41		
Shapiro–Wilk test	W = 0.9477			W = 0.9623			W = 0.9304			W = 0.9682			W = 0.9641			W = 0.9816		
Breusch-Pagan test	Pr < W = 0.6289			Pr < W = 0.8517			Pr < W = 0.6813			Pr < W = 0.5119			Pr < W = 0.7750			Pr < W = 0.9480		
Durbin–Watson test	chi2 = 2.22			chi2 = 1.63			chi2 = 3.82			chi2 = 1.94			chi2 = 3.20			chi2 = 3.17		
	Pr > chi2 = 0.357			Pr > chi2 = 0.739			Pr > chi2 = 0.550			Pr > chi2 = 0.824			Pr > chi2 = 0.419			Pr > chi2 = 0.362		
	d _L = 1.525 d _U = 1.703			d _L = 1.372 d _U = 1.808			d _L = 1.452 d _U = 1.681			d _L = 1.230 d _U = 1.786			d _L = 1.378 d _U = 1.721			d _L = 1.175 d _U = 1.854		
	D = 1.935			D = 1.877			D = 1.729			D = 1.944			D = 1.763			D = 1.921		

Table 3 – Depreciations per characteristic of HVOTL and crop.

	Ref. values		Durum wheat fields		Olive groves		Vineyards	
	Min.	Max.	Min.	Max.	Min.	Max.	Min.	Max.
Occupied area	-	-	-1.00%	-1.00%	-1.00%	-1.00%	-1.00%	-1.00%
Height (m)	18	50	-0.54%	-1.50%	-0.72%	-2.00%	-1.26%	-3.50%
Intersection (S.-W. index)	0	0.5	0.00%	-5.07%	0.00%	-8.14%	0.00%	-13.47%
Total			-1.54%	-7.57%	-1.72%	-11.14%	-2.26%	-17.97%

Figure 1 – Percentage of depreciation based on the distance from the power line.

