

# HTS\_barren\_soils\_case\_of\_latvia

May 20, 2024

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[1]: # Student: ALEKSEJS VESJOLIJS
# ST43537

# Date: 20 May 2024

# RELSTAT 2024
# Research: Hyperloop Routes Optimization Considering Barren Soil Using
↳ Operation Research, Case of Latvia

# TRANSPORTA UN SAKARU INSTITŪTS
# DATORZINĀTŅU UN TELEKOMUNIKĀCIJU FAKULTĀTE

[2]: # EXPLORATIVE DATA ANALYSIS

[3]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from pulp import LpMaximize, LpProblem, LpVariable, lpSum, LpStatus, value

[4]: # Collected data from Copernicus
data = {
    'From': ['Riga', 'Riga', 'Riga', 'Riga', 'Daugavpils', 'Daugavpils',
↳ 'Daugavpils', 'Liepaja', 'Liepaja', 'Liepaja', 'Jelgava', 'Jelgava',
↳ 'Jelgava', 'Jurmala', 'Jurmala', 'Valmiera', 'Valmiera', 'Jekabpils',
↳ 'Jelgava', 'Rezekne', 'Jekabpils', 'Liepaja', 'Jurmala', 'Ventspils',
↳ 'Jelgava', 'Ventspils', 'Valmiera', 'Rezekne', 'Jekabpils'],
    'To': ['Jelgava', 'Jurmala', 'Valmiera', 'Jekabpils', 'Jelgava', 'Rezekne',
↳ 'Jekabpils', 'Jelgava', 'Jurmala', 'Ventspils', 'Jurmala', 'Ventspils',
↳ 'Valmiera', 'Riga', 'Valmiera', 'Riga', 'Jekabpils', 'Riga', 'Daugavpils',
↳ 'Daugavpils', 'Daugavpils', 'Liepaja', 'Liepaja', 'Liepaja', 'Jelgava',
↳ 'Jurmala', 'Rezekne', 'Valmiera', 'Valmiera'],
    'Barren soil': [13, 6, 27, 29, 37, 13, 9, 45, 55, 26, 15, 40, 46, 6, 27,
↳ 29, 4, 13, 37, 13, 9, 25, 55, 26, 15, 20, 46, 24, 4],
    'Distance': [45, 38, 107, 140, 230, 89, 89, 182, 192, 118, 45, 176, 185,
↳ 38, 107, 140, 162, 45, 230, 89, 89, 182, 192, 118, 45, 154, 185, 101, 162]
}
```

```
df = pd.DataFrame(data)
```

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[5]: print(df.head())
```

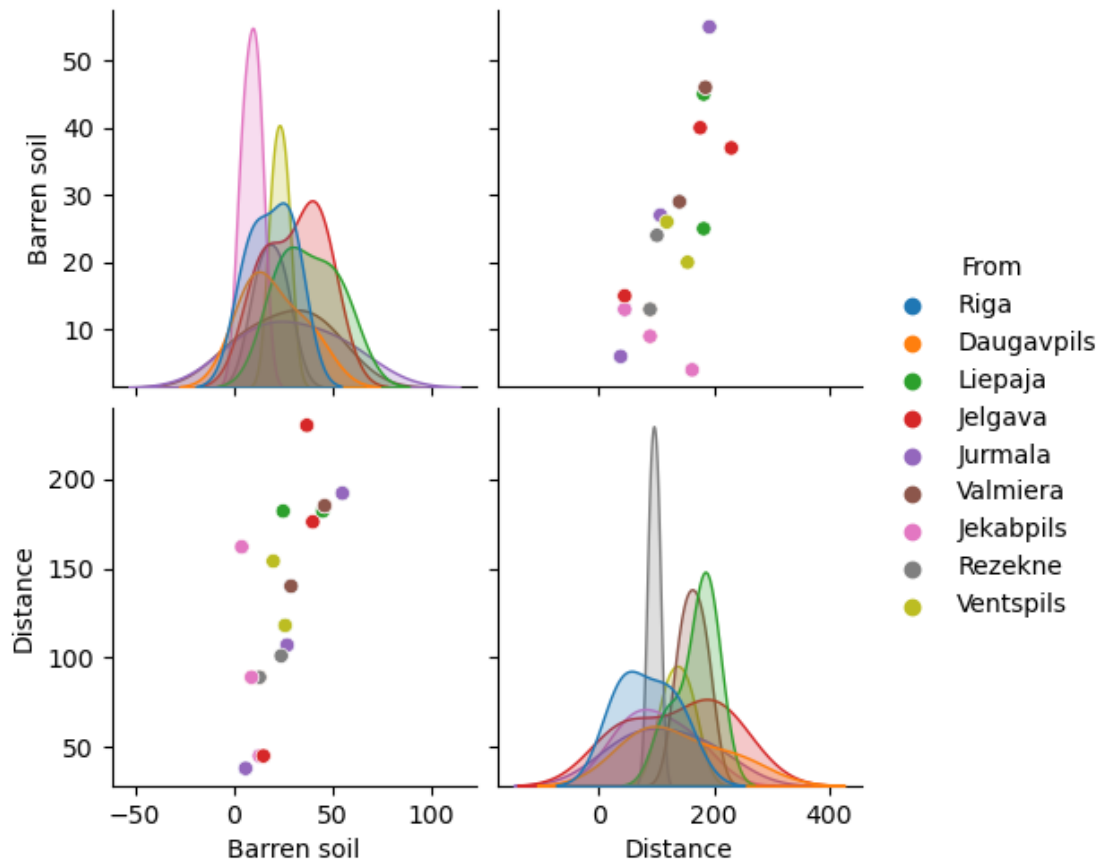
	From	To	Barren soil	Distance
0	Riga	Jelgava	13	45
1	Riga	Jurmala	6	38
2	Riga	Valmiera	27	107
3	Riga	Jekabpils	29	140
4	Daugavpils	Jelgava	37	230

```
[6]: print(df.describe())
```

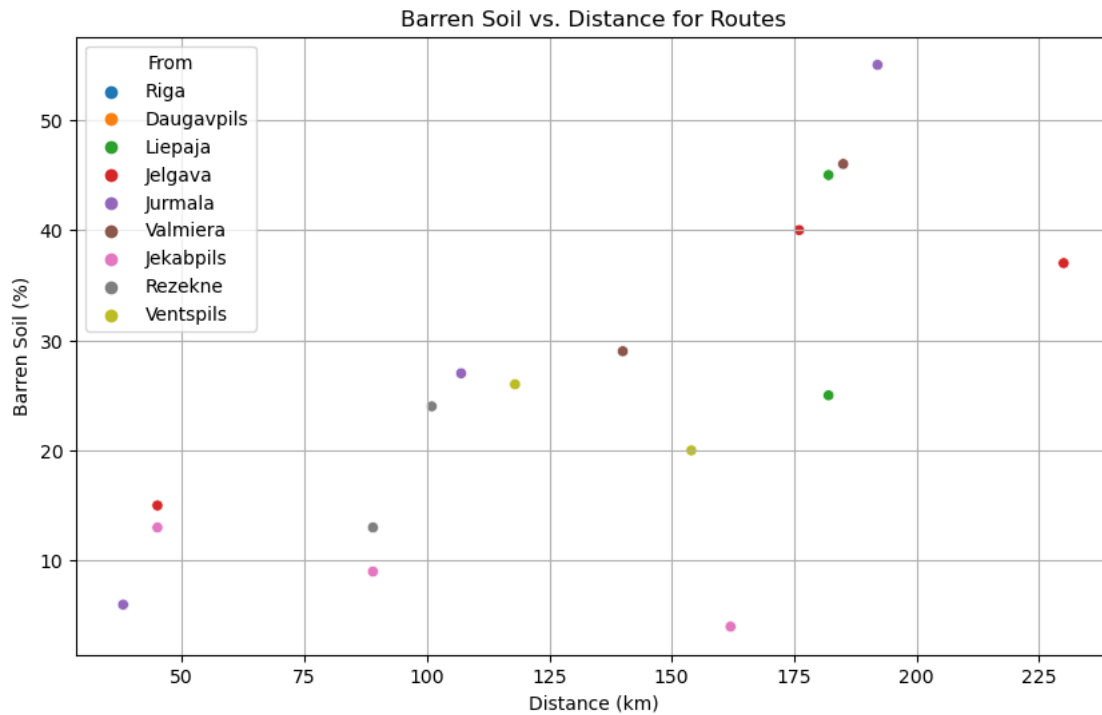
	Barren soil	Distance
count	29.00000	29.000000
mean	24.62069	126.724138
std	15.29569	59.315679
min	4.00000	38.000000
25%	13.00000	89.000000
50%	25.00000	118.000000
75%	37.00000	182.000000
max	55.00000	230.000000

```
[7]: # Create plots to visualize the data
sns.pairplot(df, hue='From')
plt.show()
```

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d:\ProgramData\anaconda3\Lib\site-packages\seaborn\_oldcore.py:1119:
FutureWarning: use_inf_as_na option is deprecated and will be removed in a
future version. Convert inf values to NaN before operating instead.
  with pd.option_context('mode.use_inf_as_na', True):
d:\ProgramData\anaconda3\Lib\site-packages\seaborn\_oldcore.py:1119:
FutureWarning: use_inf_as_na option is deprecated and will be removed in a
future version. Convert inf values to NaN before operating instead.
  with pd.option_context('mode.use_inf_as_na', True):
```



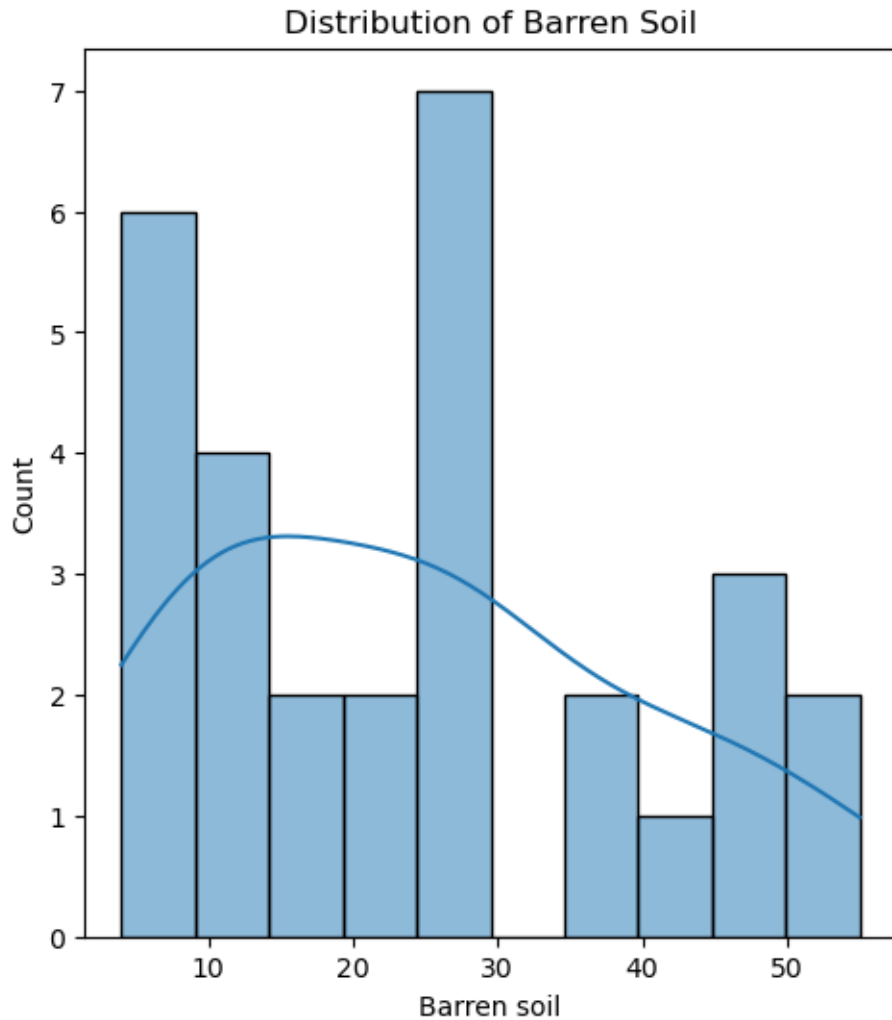
```
[8]: # Plotting Barren soil vs. Distance
plt.figure(figsize=(10, 6))
sns.scatterplot(x='Distance', y='Barren soil', hue='From', data=df)
plt.title('Barren Soil vs. Distance for Routes')
plt.xlabel('Distance (km)')
plt.ylabel('Barren Soil (%)')
plt.grid(True)
plt.show()
```



```
[9]: # Checking distribution of Barren Soil and Distance
plt.figure(figsize=(12, 6))
plt.subplot(1, 2, 1)
sns.histplot(df['Barren soil'], bins=10, kde=True)
plt.title('Distribution of Barren Soil')
```

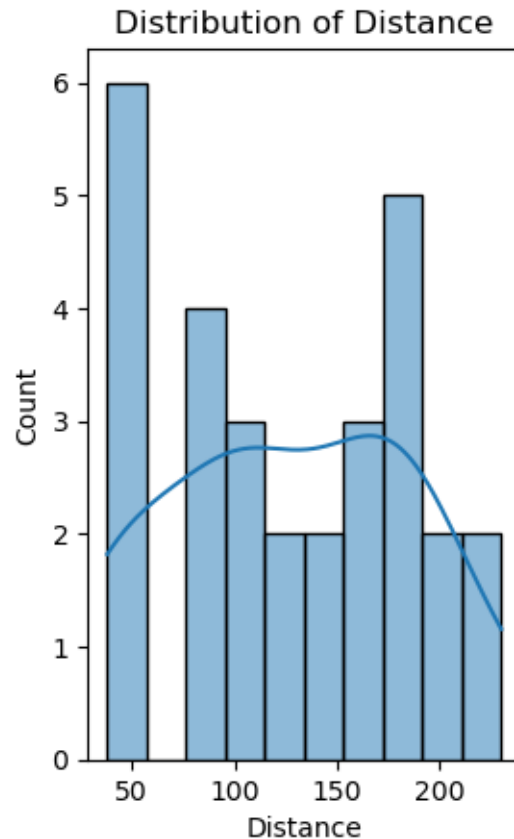
d:\ProgramData\anaconda3\Lib\site-packages\seaborn\\_oldcore.py:1119:  
FutureWarning: use\_inf\_as\_na option is deprecated and will be removed in a  
future version. Convert inf values to NaN before operating instead.  
with pd.option\_context('mode.use\_inf\_as\_na', True):

```
[9]: Text(0.5, 1.0, 'Distribution of Barren Soil')
```



```
[10]: plt.subplot(1, 2, 2)
sns.histplot(df['Distance'], bins=10, kde=True)
plt.title('Distribution of Distance')
plt.show()
```

```
d:\ProgramData\anaconda3\Lib\site-packages\seaborn\_oldcore.py:1119:
FutureWarning: use_inf_as_na option is deprecated and will be removed in a
future version. Convert inf values to NaN before operating instead.
  with pd.option_context('mode.use_inf_as_na', True):
```



```
[11]: # OPERATIONS RESEARCH PROBLEM SOLUTION
```

```
[12]: data = {
    'From': ['Riga', 'Riga', 'Riga', 'Daugavpils', 'Daugavpils', 'Liepaja', 'Jelgava', 'Jelgava', 'Valmiera', 'Jekabpils'],
    'To': ['Jelgava', 'Jurmala', 'Valmiera', 'Jelgava', 'Rezekne', 'Jelgava', 'Jurmala', 'Ventspils', 'Rezekne', 'Valmiera'],
    'Barren soil': [13, 6, 27, 37, 13, 45, 15, 40, 46, 24],
    'Distance': [45, 38, 107, 230, 89, 182, 45, 176, 185, 101]
}
df = pd.DataFrame(data)

# Create the linear programming model
model = LpProblem(name="hyperloop-route-optimization", sense=LpMaximize)

# Decision variables
route_vars = {row.Index: LpVariable(name=f"route_{row.Index}", cat='Binary')
               for row in df.itertuples()}
```

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# Objective function: Maximize the sum of barren soil * decision variable
model += lpSum(row['Barren soil'] * route_vars[idx] for idx, row in df.
    iterrows()), "Maximize_Barren_Soil_Coverage"

# Constraints: each city is connected at least once
cities = set(df['From']) | set(df['To'])
for city in cities:
    model += (lpSum(route_vars[i] for i in df.index[(df['From'] == city) |
        (df['To'] == city)]) >= 1,
        f"Connectivity_{city}")

# Solving the model
model.solve()

print("Status:", LpStatus[model.status])
print("Optimal Solution to the problem: Total Barren Soil = ", value(model.
    objective))
for var in model.variables():
    if var.value() == 1:
        print(var.name, "=", var.value())

selected_routes = df.loc[[int(v.name.split('_')[1]) for v in model.variables()
    if v.varValue == 1]]
print(selected_routes)

```

Status: Optimal

Optimal Solution to the problem: Total Barren Soil = 266.0

route\_0 = 1.0  
route\_1 = 1.0  
route\_2 = 1.0  
route\_3 = 1.0  
route\_4 = 1.0  
route\_5 = 1.0  
route\_6 = 1.0  
route\_7 = 1.0  
route\_8 = 1.0  
route\_9 = 1.0

	From	To	Barren soil	Distance
0	Riga	Jelgava	13	45
1	Riga	Jurmala	6	38
2	Riga	Valmiera	27	107
3	Daugavpils	Jelgava	37	230
4	Daugavpils	Rezekne	13	89
5	Liepaja	Jelgava	45	182
6	Jelgava	Jurmala	15	45
7	Jelgava	Ventspils	40	176
8	Valmiera	Rezekne	46	185

```
[13]: import matplotlib.pyplot as plt
import networkx as nx
import pandas as pd

# Assuming 'df' is the DataFrame and 'selected_routes' contains the routes
# ↳selected by the optimization model
data = {
    'From': ['Riga', 'Riga', 'Riga', 'Daugavpils', 'Daugavpils', 'Liepaja',
    ↳'Jelgava', 'Jelgava', 'Valmiera', 'Jekabpils'],
    'To': ['Jelgava', 'Jurmala', 'Valmiera', 'Jelgava', 'Rezekne', 'Jelgava',
    ↳'Jurmala', 'Ventspils', 'Rezekne', 'Valmiera'],
    'Barren soil': [13, 6, 27, 37, 13, 45, 15, 40, 46, 24],
    'Distance': [45, 38, 107, 230, 89, 182, 45, 176, 185, 101],
    'Selected': [1, 0, 1, 0, 1, 0, 1, 1, 0, 1] # Binary flags indicating
    ↳whether the route was selected
}
df = pd.DataFrame(data)

# Filter to only selected routes
selected_routes = df[df['Selected'] == 1]

# Create a directed graph
G = nx.DiGraph()

# Add edges based on the selected routes
for _, row in selected_routes.iterrows():
    G.add_edge(row['From'], row['To'], weight=row['Barren soil'])

# Position nodes using the spring layout
pos = nx.spring_layout(G)

# Draw the nodes
nx.draw_networkx_nodes(G, pos, node_size=700, node_color='skyblue', alpha=0.6)

# Draw the edges
nx.draw_networkx_edges(G, pos, arrowstyle='->', arrowsize=20,
    ↳edge_color='gray', width=2)

# Label the nodes
nx.draw_networkx_labels(G, pos, font_size=12, font_family='sans-serif')

# Add edge labels to show the barren soil percentage
edge_labels = nx.get_edge_attributes(G, 'weight')
nx.draw_networkx_edge_labels(G, pos, edge_labels=edge_labels, font_color='red')
```



```
plt.title('Optimized Hyperloop Routes with Barren Soil Percentage')
plt.axis('off') # Turn off the axis
plt.show()
```

### Optimized Hyperloop Routes with Barren Soil Percentage

