# **Crossover Prompt:**

### **System Prompt:**

You are an expert in the domain of optimization heuristics. Your task is to design heuristics that can effectively solve optimization problems.

Your response outputs Python code and nothing else. Format your code as a Python code string: "```python ... ```".

#### User Prompt:

Write a crossover function for Write a crossover function for solving the port selection optimization problem using a genetic algorithm. The problem requires selecting a subset of ports from a given set of ports to serve users, with the goal of minimizing the transmission power. The encoding uses binary encoding, where the number of binary bits equals the number of ports, and a bit value of 1 represents the selection of that port. The solution generated after the crossover must be valid. The 'crossover' function takes as input a 2D NumPy array parents and an integer n\_pop. The function performs a genetic crossover operation on parents to generate n\_pop offspring. Use vectorized implementation if possible.

```
[Worse code]
def crossover_v0(parents: np.ndarray, n_ pop: int) -> np.ndarray:
    n parents, n decap = parents.shape
    # Randomly select distinct parent pairs for each offspring
    parent indices = np.array([np.random.choice(n parents, size=2, replace=False)
for in range(n pop)])
    parent1 = parents[parent indices[:, 0]]
    parent2 = parents[parent indices[:, 1]]
    # Generate two random crossover points for each offspring
    cross points = np.random.randint(0, n decap, size=(n pop, 2))
    cross points.sort(axis=1)
    # Create mask for parent2 segments using broadcasting
    idx = np.arange(n decap)
    mask = (idx \ge cross points[:, 0, None]) & (idx < cross points[:, 1, None])
    # Combine genetic material using mask
    offspring = np.where(mask, parent2, parent1)
    return offspring
[Better code]
def crossover_v1(parents: np.ndarray, n_pop: int) -> np.ndarray:
```

```
n parents, n decap = parents.shape
    # Handle insufficient parents by duplicating if necessary
    if n parents == 0:
         return np.zeros((n pop, n decap), dtype=parents.dtype)
    elif n parents == 1:
         parents = np.tile(parents, (2, 1))
         n parents = 2
    # Generate parent indices for each offspring (two distinct parents per offspring)
    parent indices = np.array([np.random.choice(n parents, 2, replace=False) for
in range(n pop)])
    parent1 indices = parent indices[:, 0]
    parent2 indices = parent indices[:, 1]
    # Generate two random crossover points for each offspring and sort them
    points = np.random.randint(0, n decap + 1, size=(n pop, 2))
    points.sort(axis=1)
    # Create mask for the middle segment between crossover points
    col indices = np.arange(n decap)
    mask = (col indices >= points[:, 0, None]) & (col indices < points[:, 1, None])
    # Construct offspring: start with parent1 and replace middle segment with
parent2
    offspring = parents[parent1 indices, :].copy()
    offspring[mask] = parents[parent2 indices, :][mask]
    return offspring
[Reflection]
Here are concise design hints based on the comparison:
1. **Handle edge cases** (0-1 parents)
2. **Use direct array copying** instead of `np.where`
3. **Ensure crossover boundaries** include end indices (+1)
4. **Avoid broadcasting** where direct masking suffices
5. **Explicitly manage memory** with `.copy()`
*(18 words total)*
[Improved code]
```

Please write an improved function 'crossover\_v2', according to the reflection. Output code only and enclose your code with Python code block: '''python ... '''.

# **LLM Response:**

```
```python
import numpy as np
def crossover v2(parents: np.ndarray, n pop: int) -> np.ndarray:
     n parents, n decap = parents.shape
     # Handle edge cases: no parents or only one parent
     if n parents == 0:
         return np.zeros((n pop, n decap), dtype=parents.dtype)
     if n parents == 1:
         parents = np.vstack([parents, parents])
         n parents = 2
     # Select distinct parent pairs for each offspring
     parent indices = np.random.choice(n parents, size=(n pop, 2), replace=False)
     parent1 idx = parent indices[:, 0]
     parent2 idx = parent indices[:, 1]
     # Generate and sort crossover points
     cross points = np.random.randint(0, n decap + 1, size = (n pop, 2))
     cross points.sort(axis=1)
     # Create mask for swapping segments
```

```
col_grid = np.arange(n_decap)
mask = (col_grid >= cross_points[:, 0, None]) & (col_grid < cross_points[:, 1,
None])

# Construct offspring using segments from both parents
offspring = parents[parent1_idx].copy()
offspring[mask] = parents[parent2_idx][mask]

return offspring</pre>
```

# **Mutation Prompt:**

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#### User Prompt:

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#### [Prior reflection]

Based on prior reflections and new insights, here are concise heuristic design hints (42 words):

\*\*Prefer two-point crossover\*\* to preserve contiguous port blocks and minimize structural disruption.

```
**Enforce distinct parent pairs** maintaining diversity.
**Handle edge cases** (0-1 parents) explicitly.
**Use vectorized segment copying** instead of per-bit operations.
**Simplify with direct array assignment** avoiding complex masking.
Key focus: Structural preservation > randomization for port selection optimization.
[Code]
def crossover v1(parents: np.ndarray, n pop: int) -> np.ndarray:
     n parents, n decap = parents.shape
     if n parents == 0:
         return np.empty((n pop, n decap))
     # Generate parent indices for each offspring
     idx1 = np.random.randint(0, n parents, size=n pop)
     if n parents == 1:
         idx2 = np.zeros(n pop, dtype=int)
     else:
         idx2 = np.random.randint(0, n parents, size=n pop)
         same = idx1 == idx2
         if np.any(same):
              offset = np.random.randint(1, n parents, size=same.sum())
              idx2[same] = (idx2[same] + offset) \% n parents
     # Generate two crossover points for each offspring and sort them
     if n decap > 0:
         cross points = np.random.randint(0, n decap, size=(n pop, 2))
         cross points.sort(axis=1)
         # Create mask for the segment between crossover points
         col indices = np.arange(n decap)
         mask = (col indices >= cross points[:, 0, None]) & (col indices <
cross points[:, 1, None])
         # Construct offspring using segments from both parents
         offspring = np.where(mask, parents[idx2], parents[idx1])
     else:
         offspring = parents[idx1]
     return offspring
LLM Response:
```python
```

import numpy as np

```
def crossover v2(parents: np.ndarray, n pop: int) -> np.ndarray:
    n parents, n decap = parents.shape
    if n parents == 0:
         return np.empty((n_pop, n_decap))
    # Generate distinct parent pairs for each offspring
    idx1 = np.random.randint(0, n parents, size=n pop)
    if n parents == 1:
         idx2 = np.zeros(n pop, dtype=int)
    else:
         idx2 = np.random.randint(0, n parents, size=n pop)
         same = idx1 == idx2
         if np.any(same):
              offset = np.random.randint(1, n parents, size=same.sum())
              idx2[same] = (idx2[same] + offset) \% n parents
    # Handle crossover when ports exist
    if n decap > 0:
         # Generate crossover points in [0, n_decap+1] to include full segment
possibility
         cross points = np.random.randint(0, n decap+1, size=(n pop, 2))
         cross_points.sort(axis=1)
         # Vectorized mask creation for segment between points
         col indices = np.arange(n decap)
         mask = (col indices >= cross points[:, 0, None]) & (col indices <
```

```
cross_points[:, 1, None])

# Construct offspring by combining parent segments
    offspring = np.where(mask, parents[idx2], parents[idx1])
else:
    offspring = parents[idx1]

return offspring
```