

```
In [6]: #primal
import numpy as np
from scipy.optimize import linprog
primal_c = [-3, -5]
primal_A = [[1, 1],
            [1, 3]]
primal_b = [6, 9]
bounds = [(0, None), (0, None)]
primal = linprog(primal_c,
                  A_ub=primal_A,
                  b_ub=primal_b,
                  bounds=bounds,
                  method="highs")
print(primal)

message: Optimization terminated successfully. (HiGHS Status 7: Optimal)
success: True
status: 0
fun: -21.0
x: [ 4.500e+00  1.500e+00]
nit: 2
lower: residual: [ 4.500e+00  1.500e+00]
      marginals: [ 0.000e+00  0.000e+00]
upper: residual: [          inf          inf]
      marginals: [ 0.000e+00  0.000e+00]
eqlin: residual: []
      marginals: []
ineqlin: residual: [ 0.000e+00  0.000e+00]
        marginals: [-2.000e+00 -1.000e+00]
mip_node_count: 0
mip_dual_bound: 0.0
mip_gap: 0.0
```

```
In [4]: print(primal.x)
```

```
[4.5 1.5]
```

```
In [5]: print(-primal.fun)
```

```
21.0
```

```
In [9]: print(primal.ineqlin.marginals)
```

```
[-2. -1.]
```

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In [8]: #dual
import numpy as np
from scipy.optimize import linprog
dual_c = [6, 9]
dual_A = [[-1, -1],
          [-1, -3]]
dual_b = [-3, -5]
bounds = [(0, None), (0, None)]
dual = linprog(dual_c,
                A_ub=dual_A,
                b_ub=dual_b,
                bounds=bounds,
                method="highs")
print(dual)
```

```

message: Optimization terminated successfully. (HiGHS Status 7: Optimal)
success: True
status: 0
  fun: 21.0
   x: [ 2.000e+00  1.000e+00]
  nit: 2
lower: residual: [ 2.000e+00  1.000e+00]
      marginals: [ 0.000e+00  0.000e+00]
upper: residual: [          inf          inf]
      marginals: [ 0.000e+00  0.000e+00]
eqlin: residual: []
      marginals: []
ineqlin: residual: [ 0.000e+00  0.000e+00]
        marginals: [-4.500e+00 -1.500e+00]
mip_node_count: 0
mip_dual_bound: 0.0
  mip_gap: 0.0

```

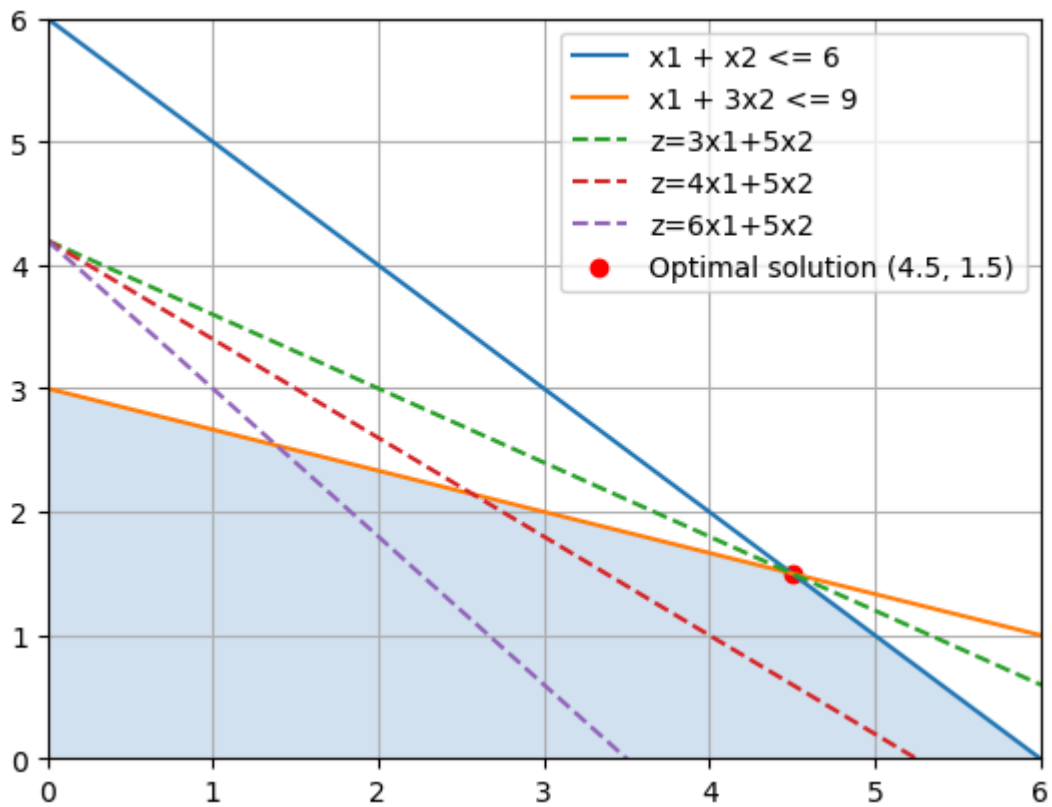
```
In [10]: print(dual.ineqlin.marginals)
```

```
[-4.5 -1.5]
```

```
In [17]: # shadow primal
import matplotlib.pyplot as plt
import numpy as np
x = np.linspace(0, 7, 400)
constraint_1 = 6 - x
constraint_2 = (9 - x) / 3
z_original=(21-3*x)/5
z_small=(21-4*x)/5
z_large=(21-6*x)/5
plt.figure()
plt.plot(x, constraint_1, label="x1 + x2 <= 6")
plt.plot(x, constraint_2, label="x1 + 3x2 <= 9")
plt.plot(x,z_original,"--",label="z=3x1+5x2")
plt.plot(x,z_small,"--",label="z=4x1+5x2")
plt.plot(x,z_large,"--",label="z=6x1+5x2")
plt.fill_between(x, 0, np.minimum(constraint_1, constraint_2), alpha=0.2)
plt.scatter(primal.x[0], primal.x[1], color='red', label="Optimal solution (4.5,

plt.xlim(0, 6)
plt.ylim(0, 6)

plt.legend()
plt.grid()
plt.show()
```



```
In [5]: import numpy as np
from scipy.optimize import linprog
def primal_dual(C, A, b):
    A = np.array(A)
    b = np.array(b)
    c = np.array(C)

    primal = linprog(
        c=-c,
        A_ub=A,
        b_ub=b,
        bounds=[(0, None)] * len(c),
        method='highs'
    )

    dual = linprog(
        c=b,
        A_ub=-A.T,
        b_ub=-c,
        bounds=[(0, None)] * len(b),
        method='highs'
    )

    print(primal.x)
    print(primal.fun)
    print(dual.x)
C = [3, 5]
A = [[1, 1], [1, 3]]
B = [6, 9]
primal_dual(C, A, B)
```

```
[4.5 1.5]
-21.0
[2. 1.]
```