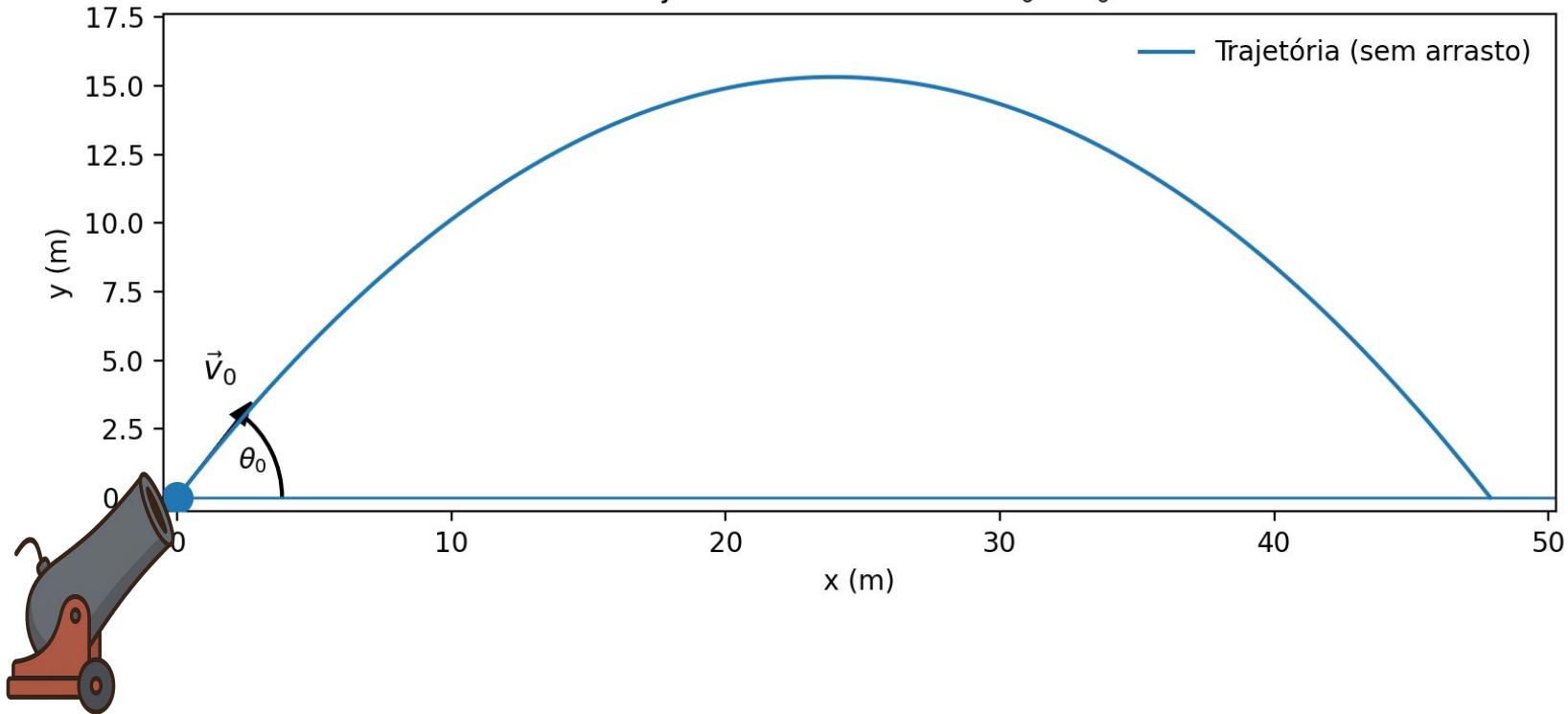




DISPARANDO  
CANHÕES...

...usando PINN

### Trajetória balística com $\vec{v}_0$ e $\theta_0$



# FÍSICA DO DISPARO BALÍSTICO...

...COM ARRASTO! (MAS SEM VENTO)

$$\dot{x} = v_x$$

$$\dot{y} = v_y$$

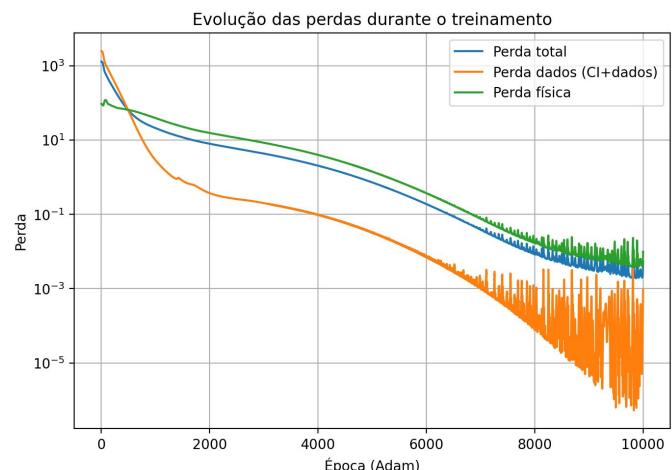
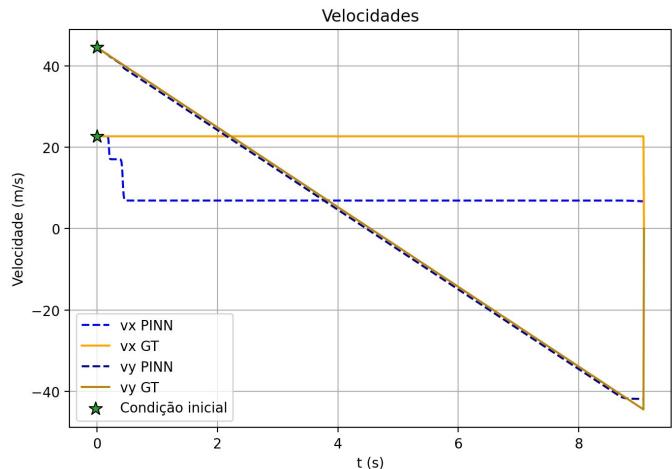
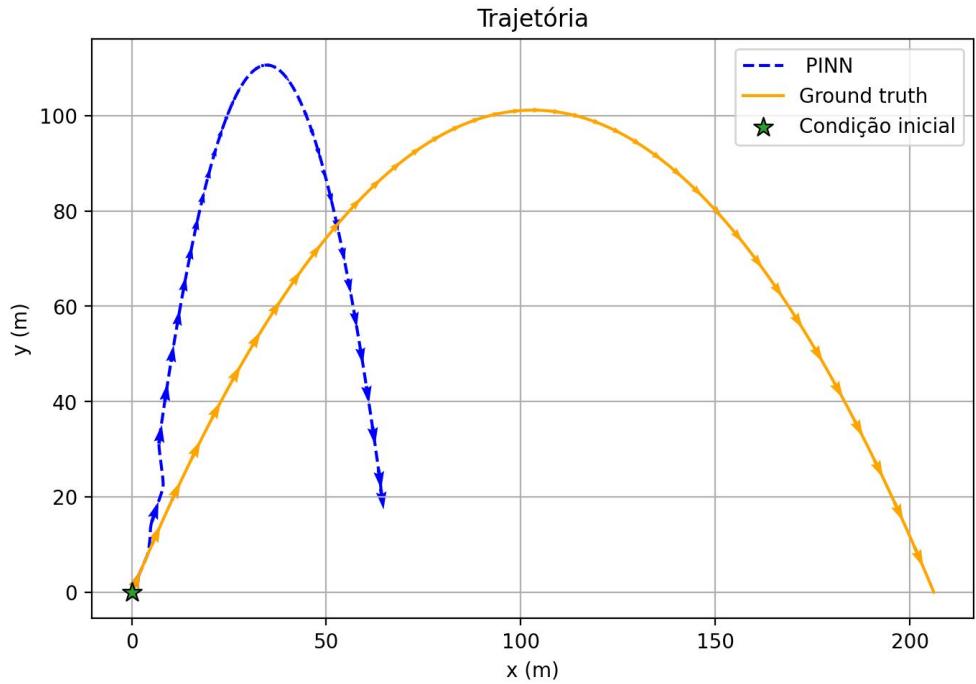
$$\dot{v}_x = -k |\vec{v}| v_x$$

$$\dot{v}_y = -g - k |\vec{v}| v_y$$

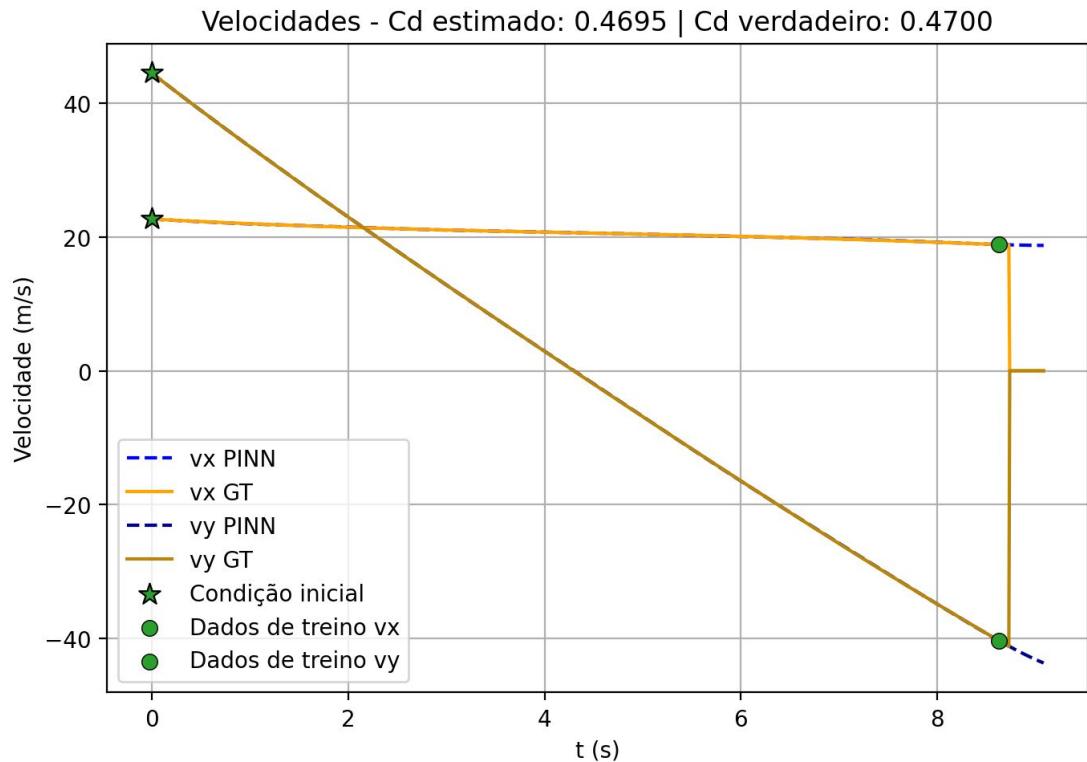
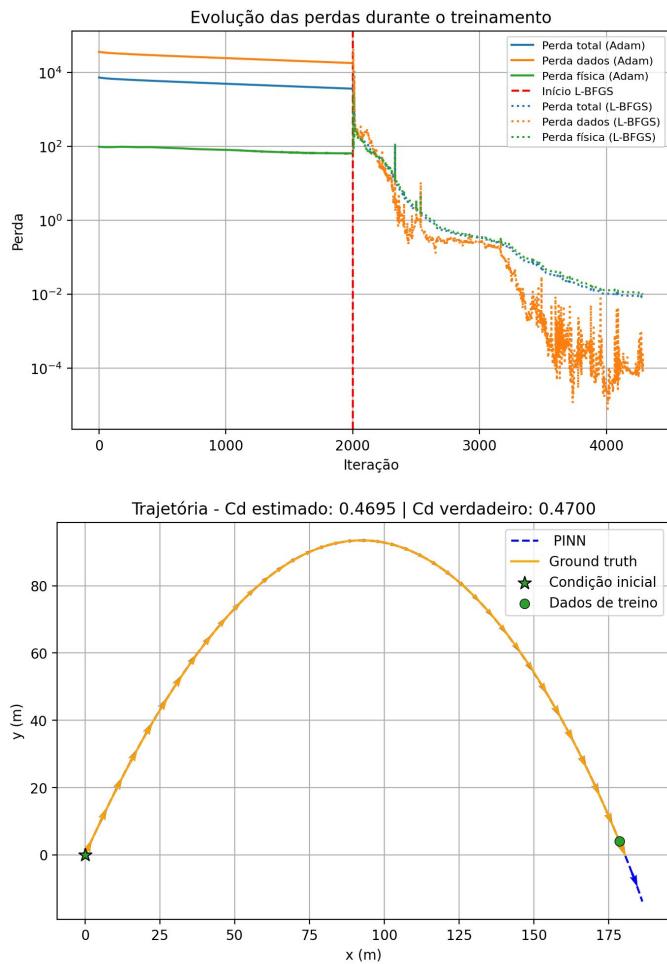
onde  $k = \frac{\rho C_d A}{2m}$

e  $|\vec{v}| = \sqrt{v_x^2 + v_y^2}$

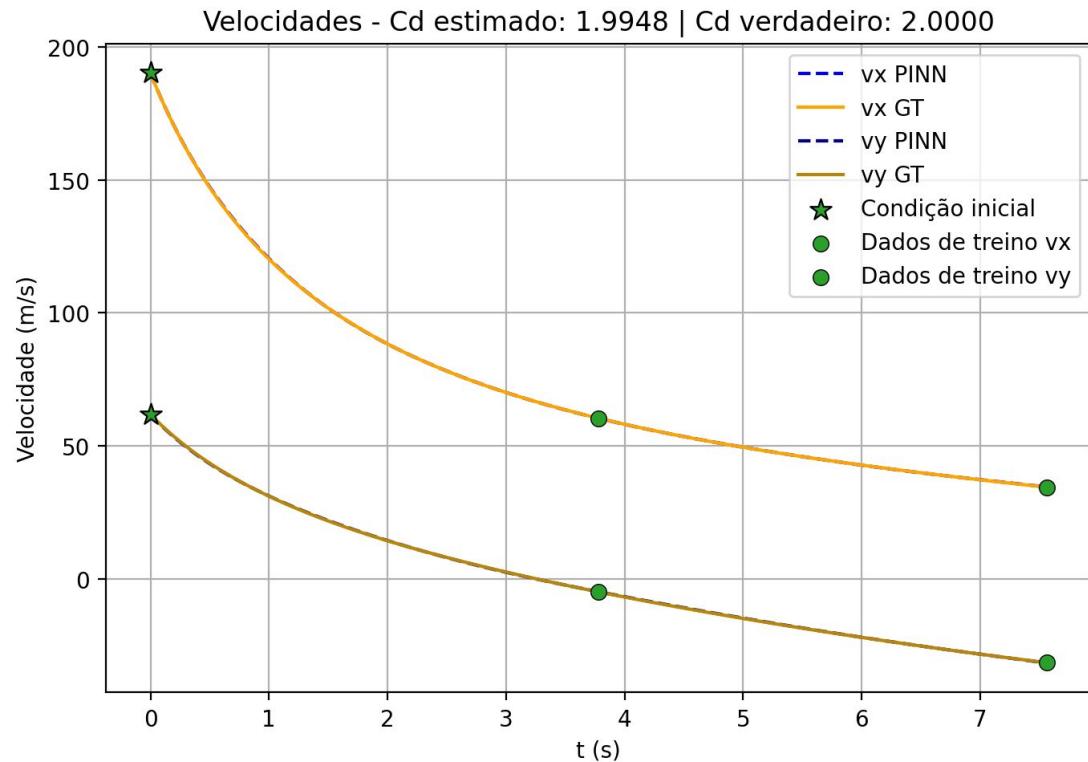
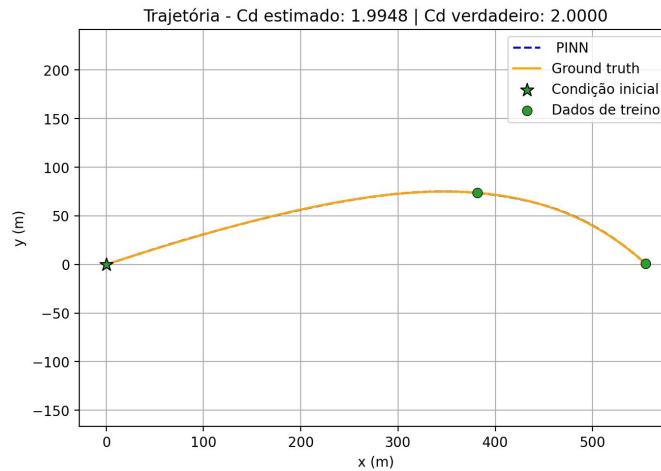
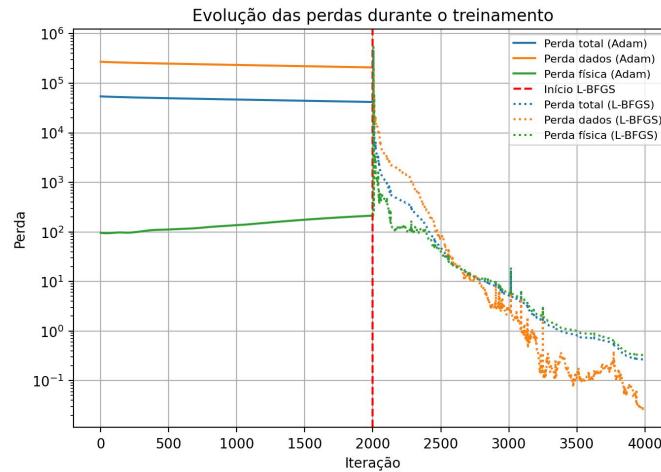
Rede [1-64-64-64-4], Adam\_steps=10000, N\_phys=500



# Adicionando apenas 1 dado extra



# Mudando as condições iniciais - Tiro mais rasante, com alto arrasto



# PRÓXIMOS PASSOS??

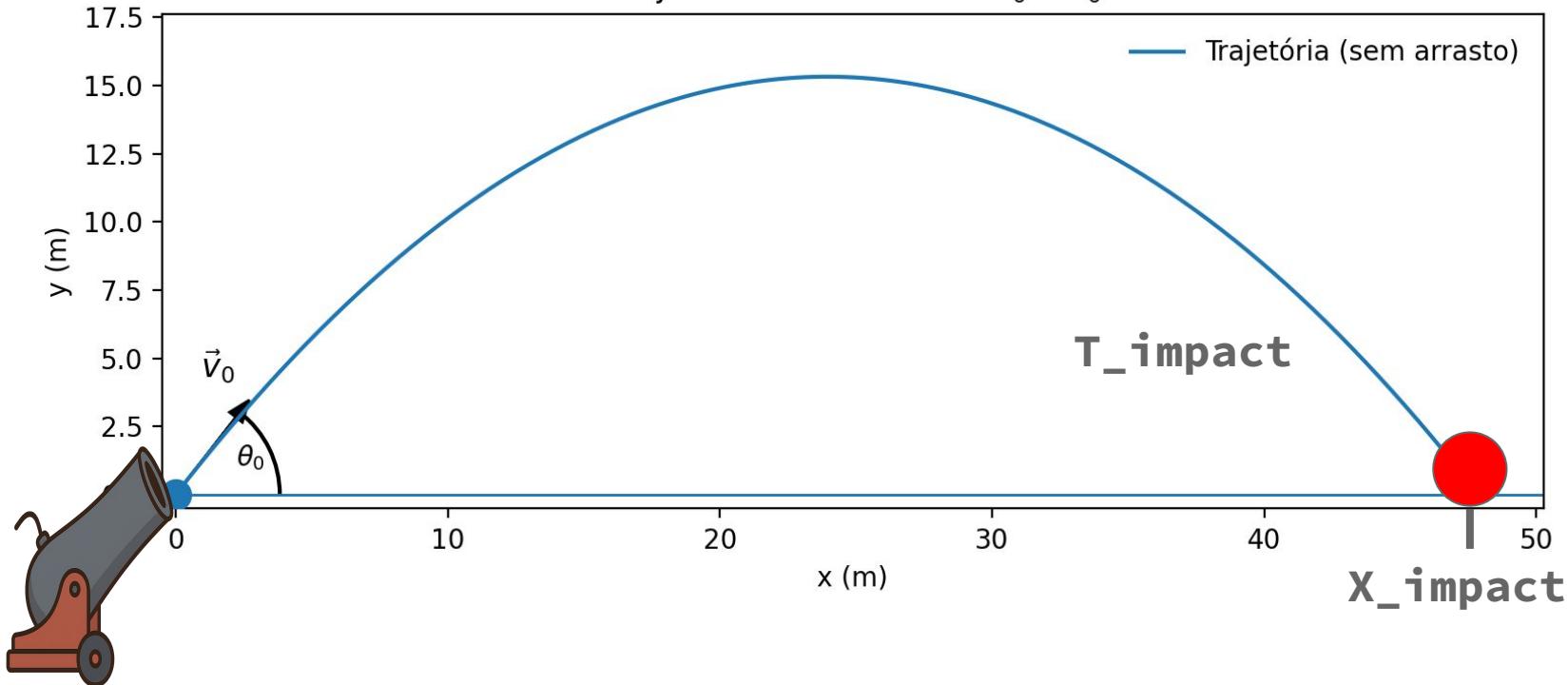
USAR PINN PARA  
ESTIMAR O ALVO, NÃO A  
TRAJETÓRIA

# REDE NEURAL QUE:

Entrada      ( **v<sub>θ</sub>** ; **θ<sub>θ</sub>** )

Saída      ( **x\_impact** ; **T\_impact** )

### Trajetória balística com $\vec{v}_0$ e $\theta_0$



# FÍSICA - EQUAÇÕES INTEGRAIS - VÁCUO

$$x(T) = \int_0^T v_x(t) dt = \int_0^T v_0 \cos \theta dt$$

$$y(T) = \int_0^T v_y(t) dt = \int_0^T (v_0 \sin \theta - gt) dt$$

# FÍSICA - CONTORNOS

$v_y(T_{\text{impact}}) < 0 \quad \text{(evita solução trivial)} \quad T=0$

$$y(T_{\text{impact}}) = 0$$

$$x(T_{\text{impact}}) = x_{\text{impact}}$$

$$v_y(T_{\text{impact}}) < 0$$

```
raw = self.net(inputs)

# T_impact deve ser positivo (usamos softplus para garantir)
T_impact = torch.nn.functional.softplus(raw[:, 0:1]) + 0.1 # mínimo de 0.1s

# x_impact deve ser positivo
x_impact = torch.nn.functional.softplus(raw[:, 1:2])

vx = vx0.unsqueeze(1).expand(-1, self.n_integration) # (batch, n_integration)
vy = vy0.unsqueeze(1) - self.g * t # (batch, n_integration)

# Integração por trapézio
#  $x = \int vx \, dt$ 
x_integrated = torch.trapezoid(vx, t, dim=1) # (batch,)

#  $y = \int vy \, dt$ 
y_integrated = torch.trapezoid(vy, t, dim=1) # (batch,)

# vy no momento do impacto
vy_impact = vy0 - self.g * T_impact.squeeze(1) # (batch,)
```

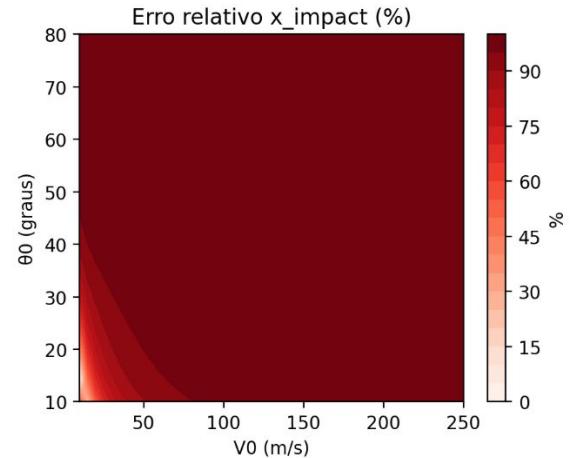
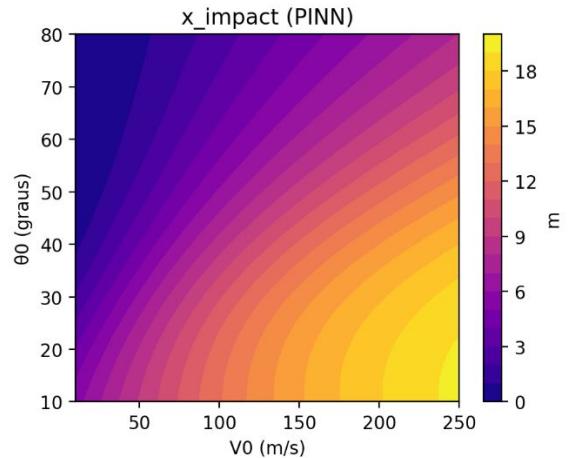
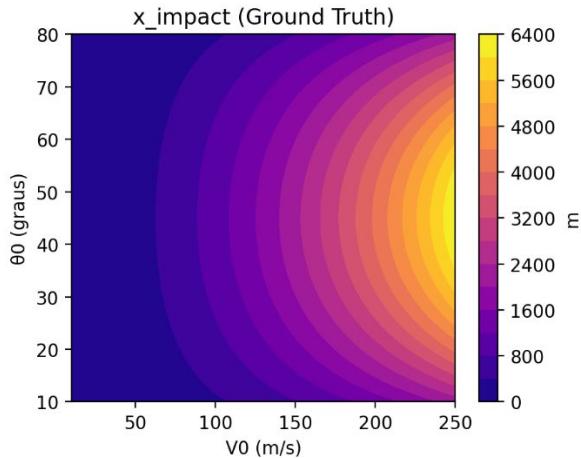
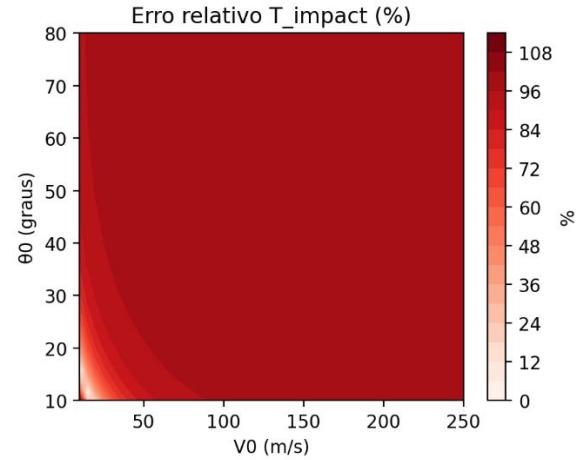
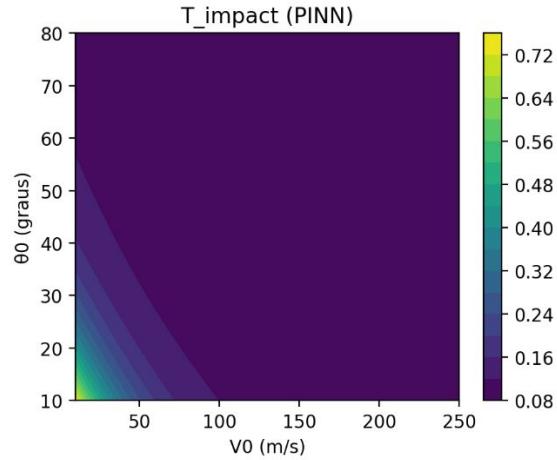
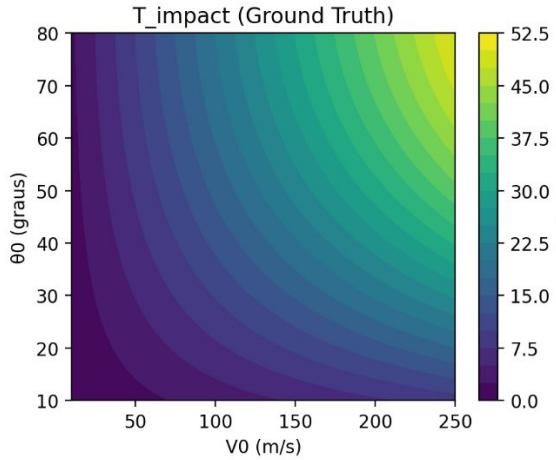
```
# Loss 1:  $y(T_{impact}) = \theta$ 
loss_y = (y_integrated ** 2).mean()

# Loss 2:  $x_{integrado} = x_{impact}$  predito
loss_x = ((x_integrated - x_impact.squeeze(1)) ** 2).mean()

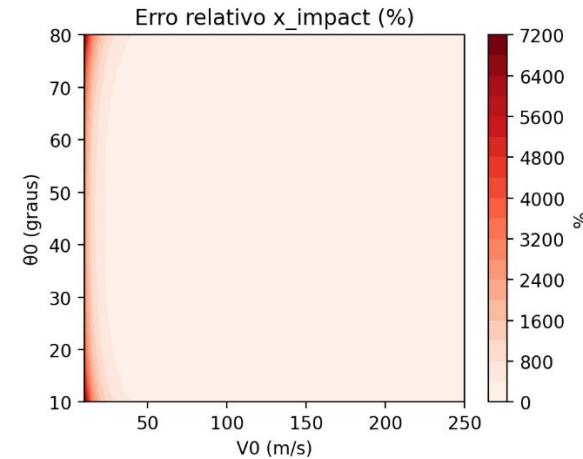
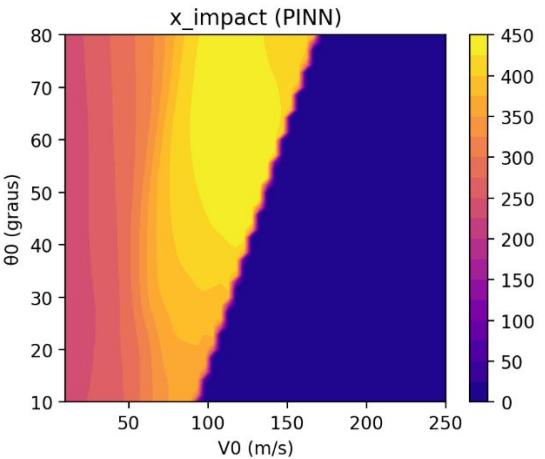
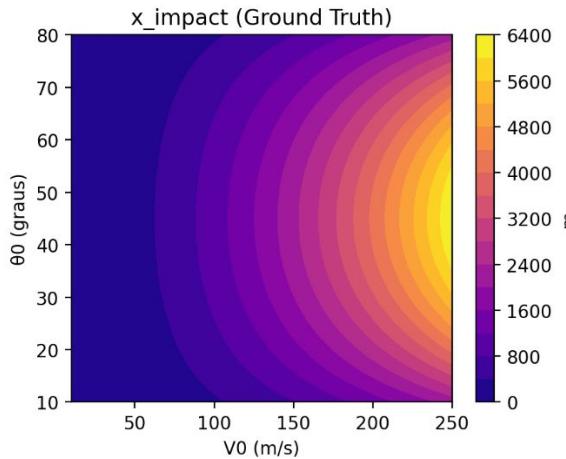
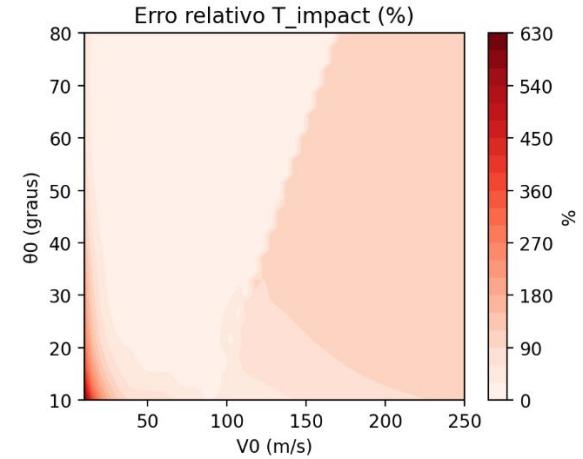
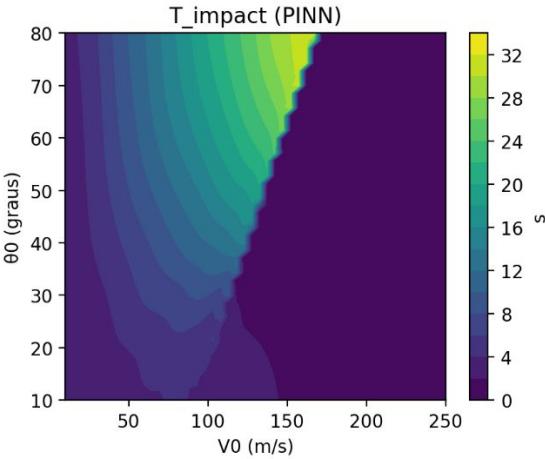
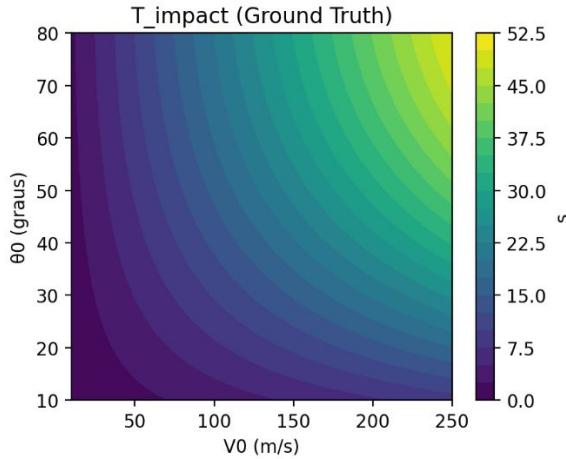
# Loss 3:  $vy_{impact} < \theta$  (penaliza se  $vy \geq \theta$ )
# Usamos ReLU:  $\max(\theta, vy_{impact} + \epsilon_{vy})$  para margem de segurança
loss_vy = torch.relu(vy_impact + self.epsilon_vy).mean()

# Perda total com pesos configuráveis
loss = self.lambda_y * loss_y + self.lambda_x * loss_x + self.lambda_vy * loss_vy
```

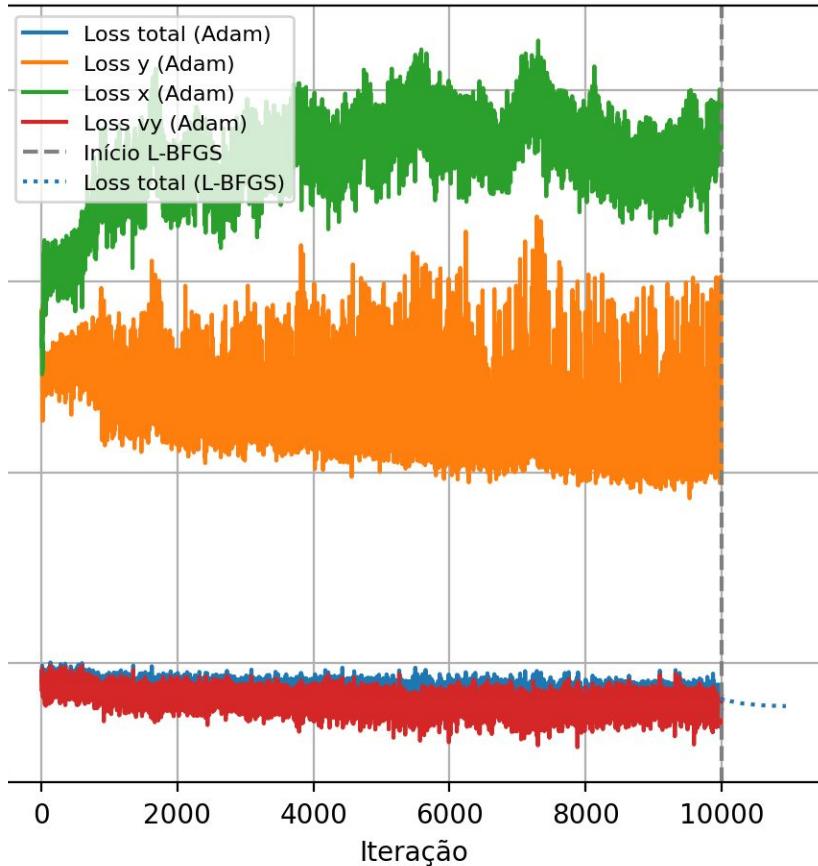
Comparação PINN vs Ground Truth  
RMSE T: 21.2480s | RMSE X: 2315.3138m



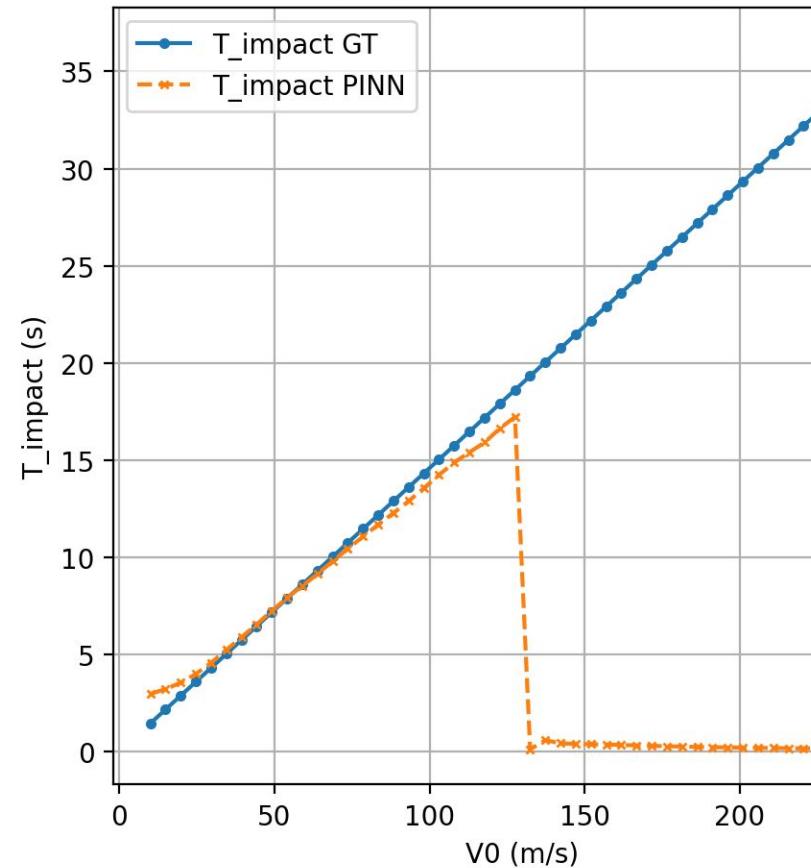
Comparação PINN vs Ground Truth  
RMSE T: 18.9469s | RMSE X: 2294.2887m



## Evolução das perdas



## T\_impact vs V0 ( $\theta \approx 46^\circ$ )

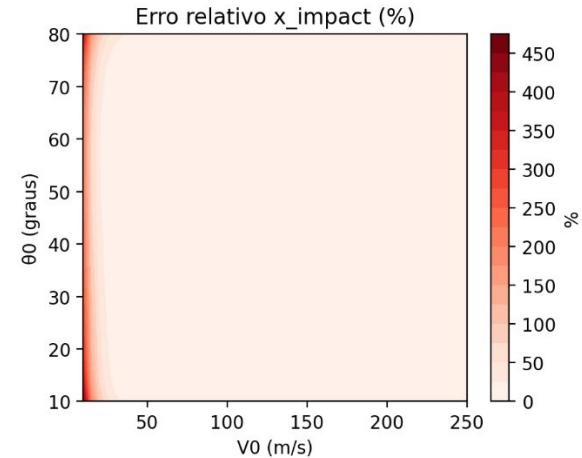
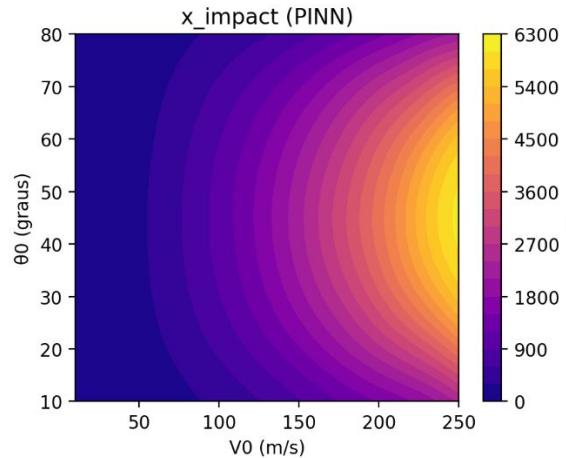
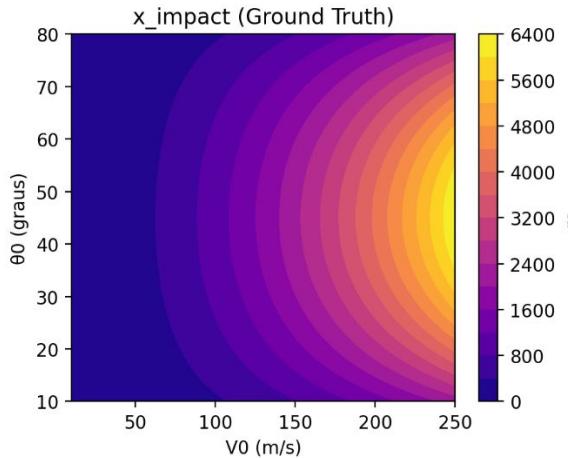
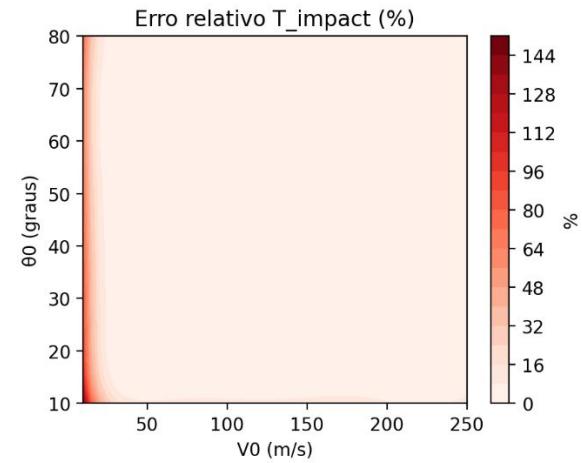
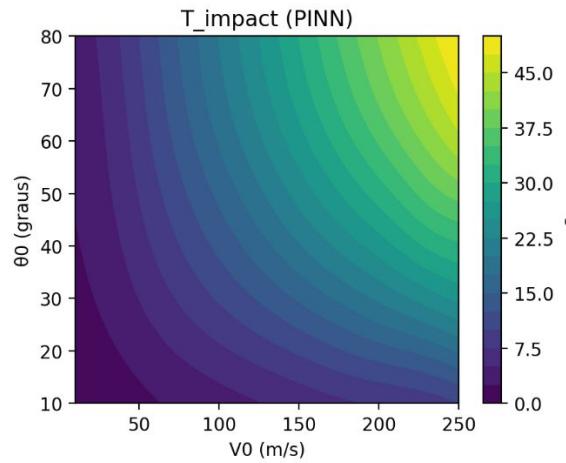
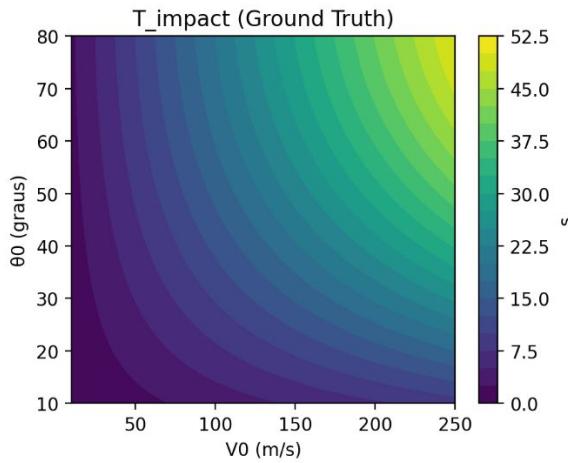


```
# Escalas esperadas para outputs (baseadas na física)
#  $T_{max} \approx 2 * V_0_{max} * \sin(\theta_{0_{max}}) / g$ 
T_scale = 2 * V0_max * np.sin(np.radians(theta0_max)) / g # ≈ 50s
#  $x_{max} \approx V_0_{max}^2 / g$  (para  $\theta=45^\circ$ )
X_scale = V0_max**2 / g # ≈ 6370m

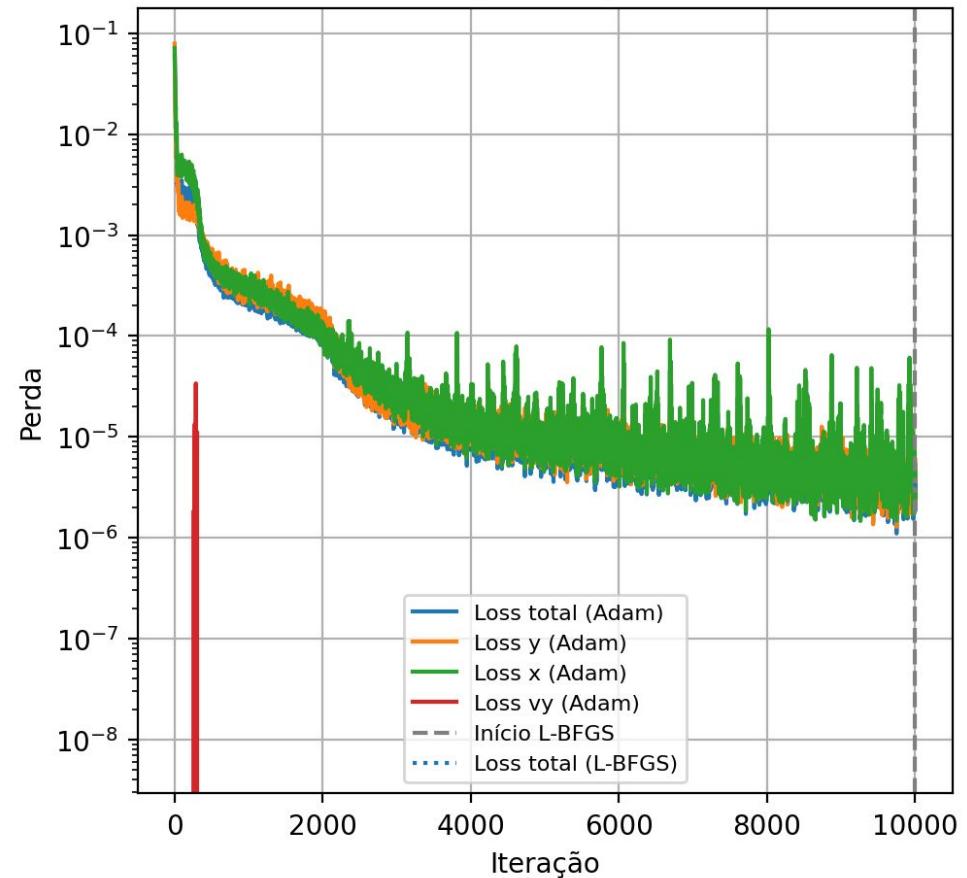
#  $T_{impact}$  deve ser positivo (usamos softplus para garantir)
T_impact = torch.nn.functional.softplus(raw[:, 0:1]) + 0.1 # mínimo de 0.1s
#  $T_{impact}$ : sigmoid * T_scale (range: [0, T_scale])
# Adicionamos pequeno offset para evitar T=0
T_impact = torch.sigmoid(raw[:, 0:1]) * self.T_scale + 0.1

#  $x_{impact}$  deve ser positivo
x_impact = torch.nn.functional.softplus(raw[:, 1:2])
#  $x_{impact}$ : sigmoid * X_scale (range: [0, X_scale])
x_impact = torch.sigmoid(raw[:, 1:2]) * self.X_scale
```

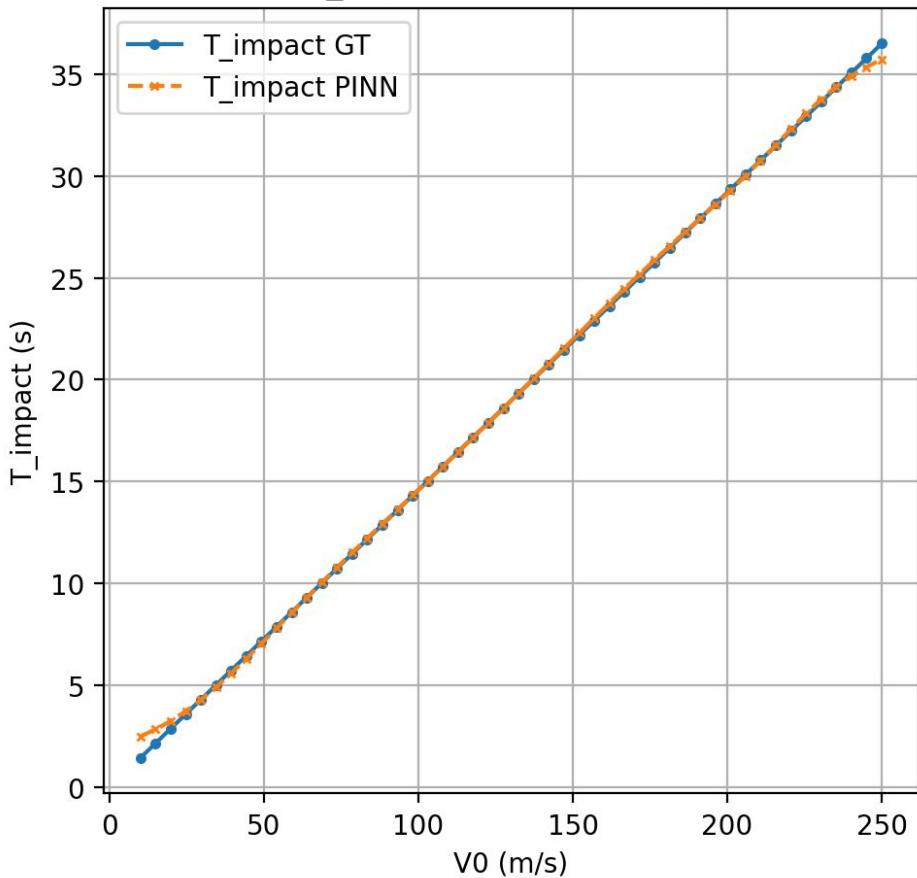
Comparação PINN vs Ground Truth  
RMSE T: 0.2261s | RMSE X: 29.1307m



### Evolução das perdas



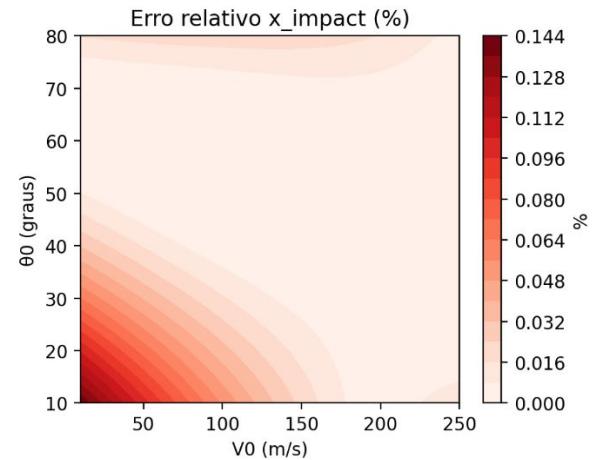
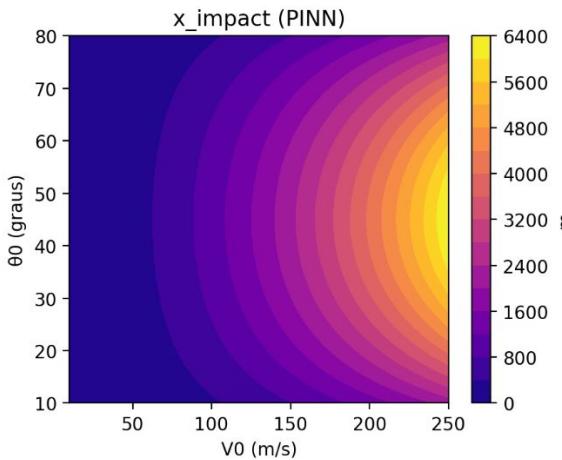
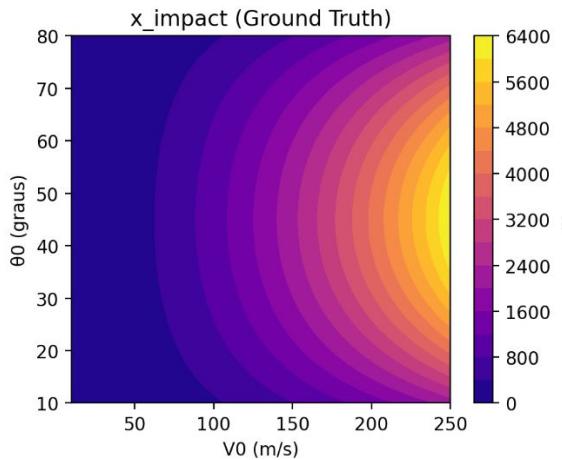
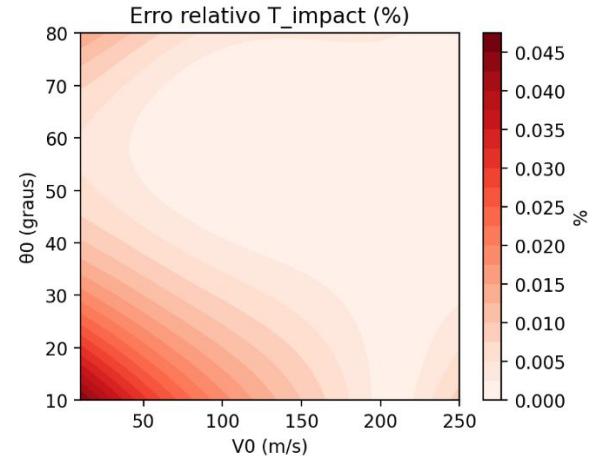
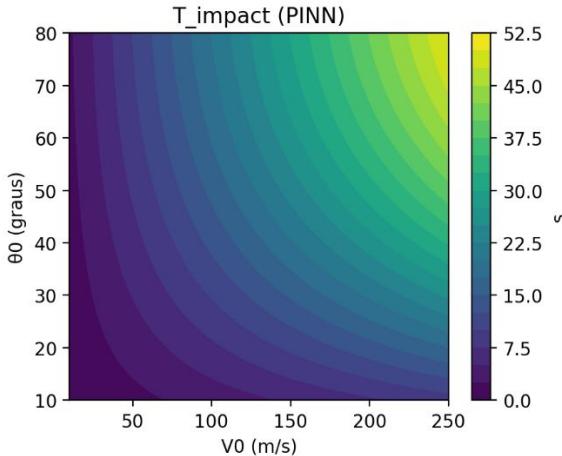
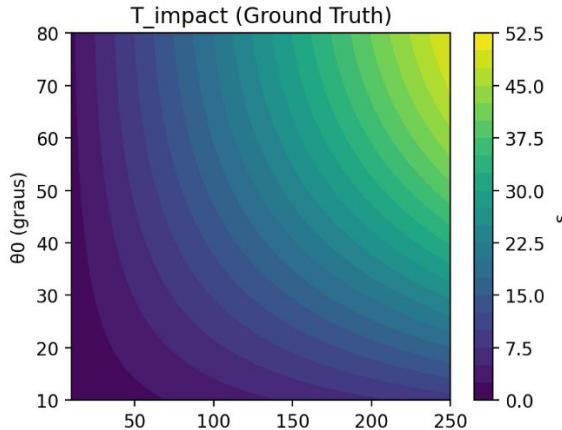
### T\_impact vs V0 ( $\theta \approx 46^\circ$ )

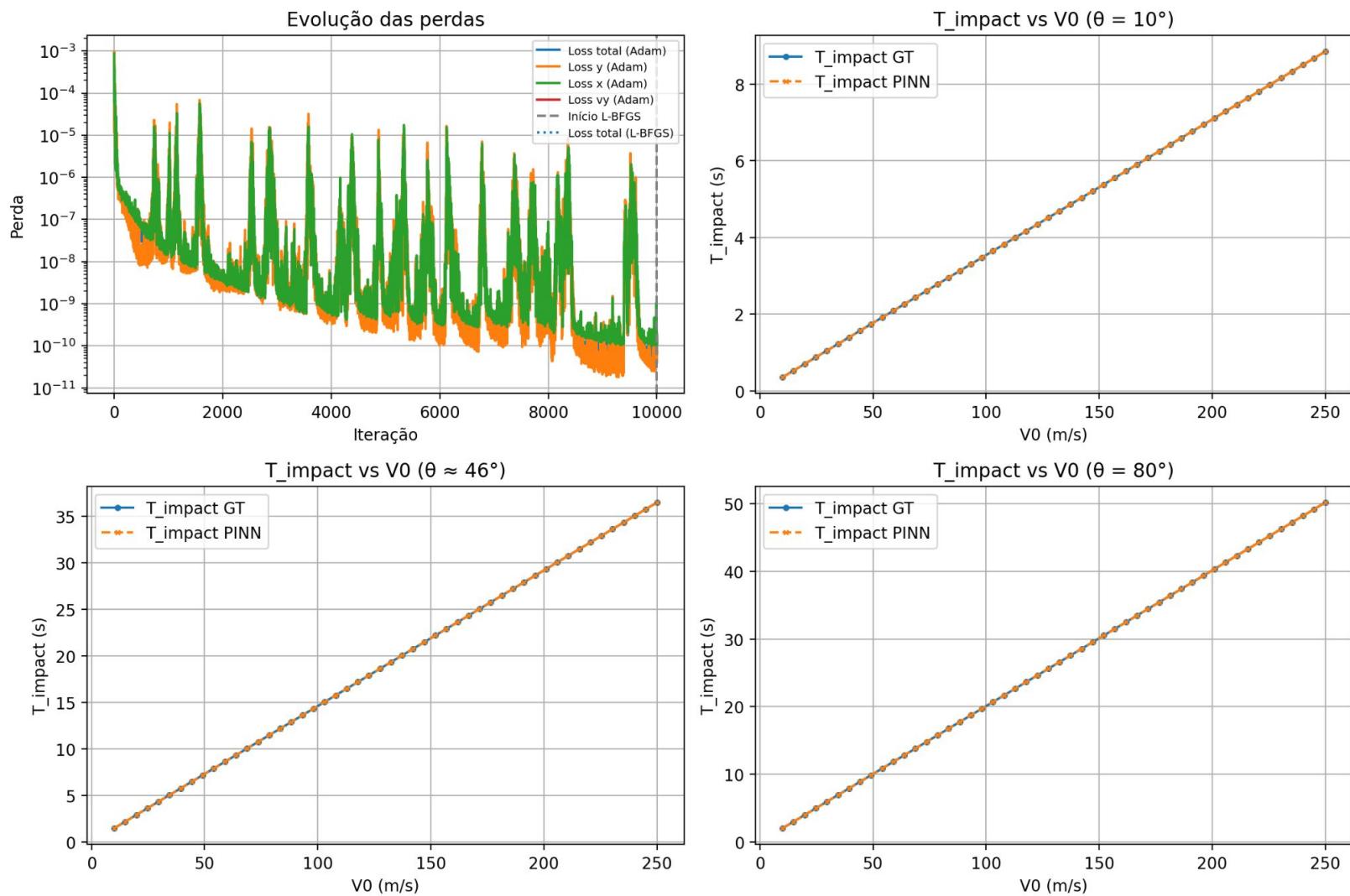


```
# Escala esperada baseada na física (vácuo)
# T_expected = 2 * vθ * sin(thetaθ) / g
# X_expected = vθ² * sin(2*thetaθ) / g
T_expected = 2 * vθ * torch.sin(thetaθ) / self.g
X_expected = vθ**2 * torch.sin(2 * thetaθ) / self.g

# Rede prediz fator de ajuste: sigmoid dá [0, 1], multiplicamos por 2 para [0, 2x esperado]
# Isso garante que a sigmoid sempre opera numa faixa útil
T_impact = torch.sigmoid(raw[:, 0:1]) * 2 * T_expected
x_impact = torch.sigmoid(raw[:, 1:2]) * 2 * X_expected
```

Comparação PINN vs Ground Truth  
RMSE T: 0.0004s | RMSE X: 0.1022m





Checkpoint salvo em D:\Code\cannon\_shot\_pinn\checkpoints\impact\_vacuo\impact-2\_64\_64\_64  
Melhor loss: 5.734229e-11 no passo 9374

Treino finalizado.

✓ # ===== Usar melhor modelo ===== ...

... Usando melhor modelo (passo 9374) para avaliação

✓ # ===== Avaliação ===== ...

RMSE T\_impact: 4.3421e-04 s

RMSE x\_impact: 1.0216e-01 m

Erro relativo médio T: 0.01%

Erro relativo médio X: 0.01%

# FÍSICA - EQUAÇÕES INTEGRAIS - ARRASTO

$$x(T) = \int_0^T v_x(t) dt$$

$$y(T) = \int_0^T v_y(t) dt$$

$$\begin{aligned}\dot{v}_x &= -k |\vec{v}| v_x \\ \dot{v}_y &= -g - k |\vec{v}| v_y\end{aligned}$$

onde  $k = \frac{\rho C_d A}{2m}$

$$\text{e } |\vec{v}| = \sqrt{v_x^2 + v_y^2}$$

```
def compute_physics(self, v0, theta0_rad, T_impact):
    """
    Calcula y_impact e x_integrated usando integração numérica (Euler) com arrasto.

    v0: tensor (batch,) - velocidade inicial
    theta0_rad: tensor (batch,) - ângulo em radianos
    T_impact: tensor (batch, 1) - tempo de impacto predito

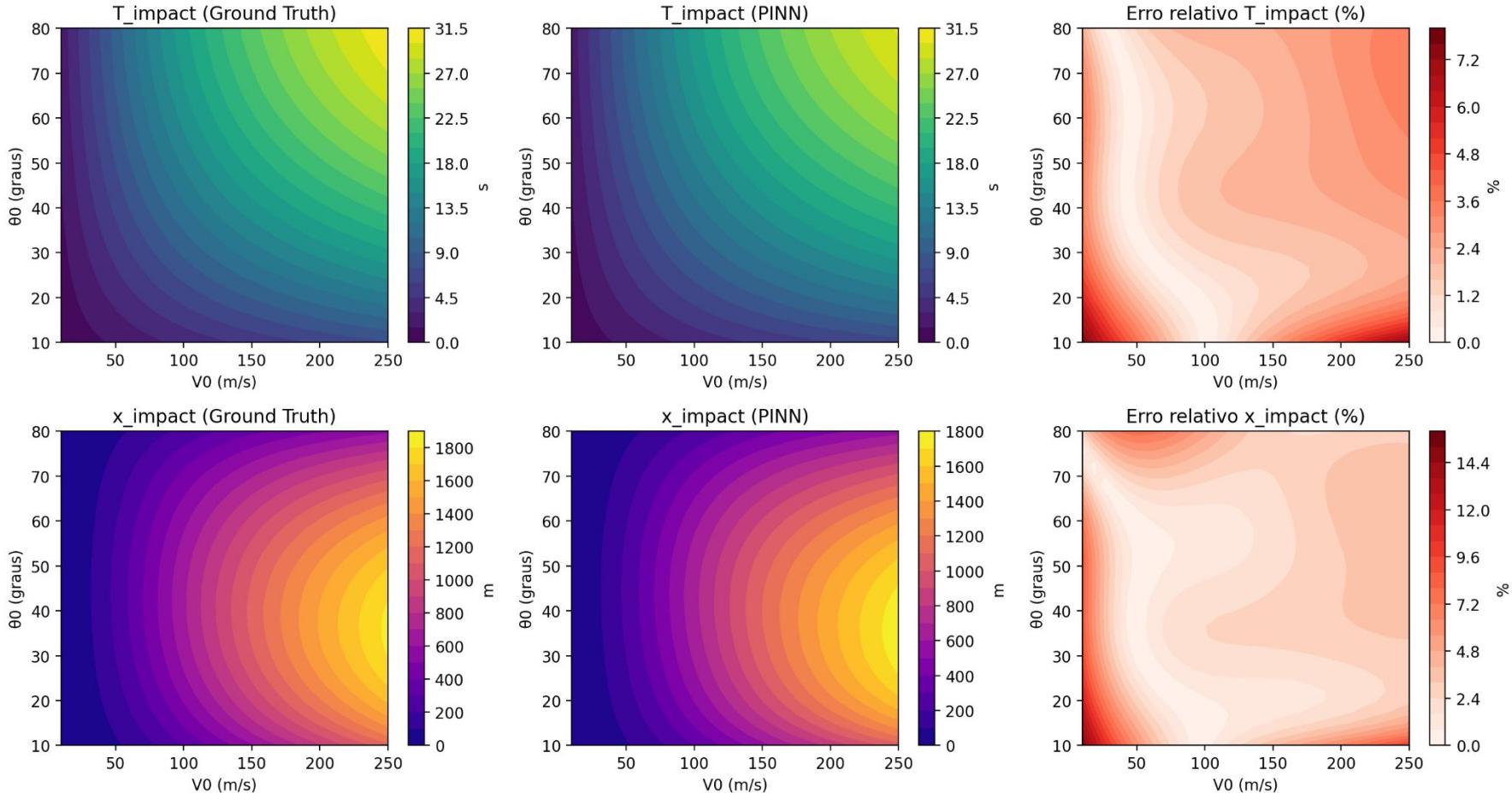
    Retorna: y_final, x_final, vy_final
    """
    batch_size = v0.shape[0]
    device = v0.device

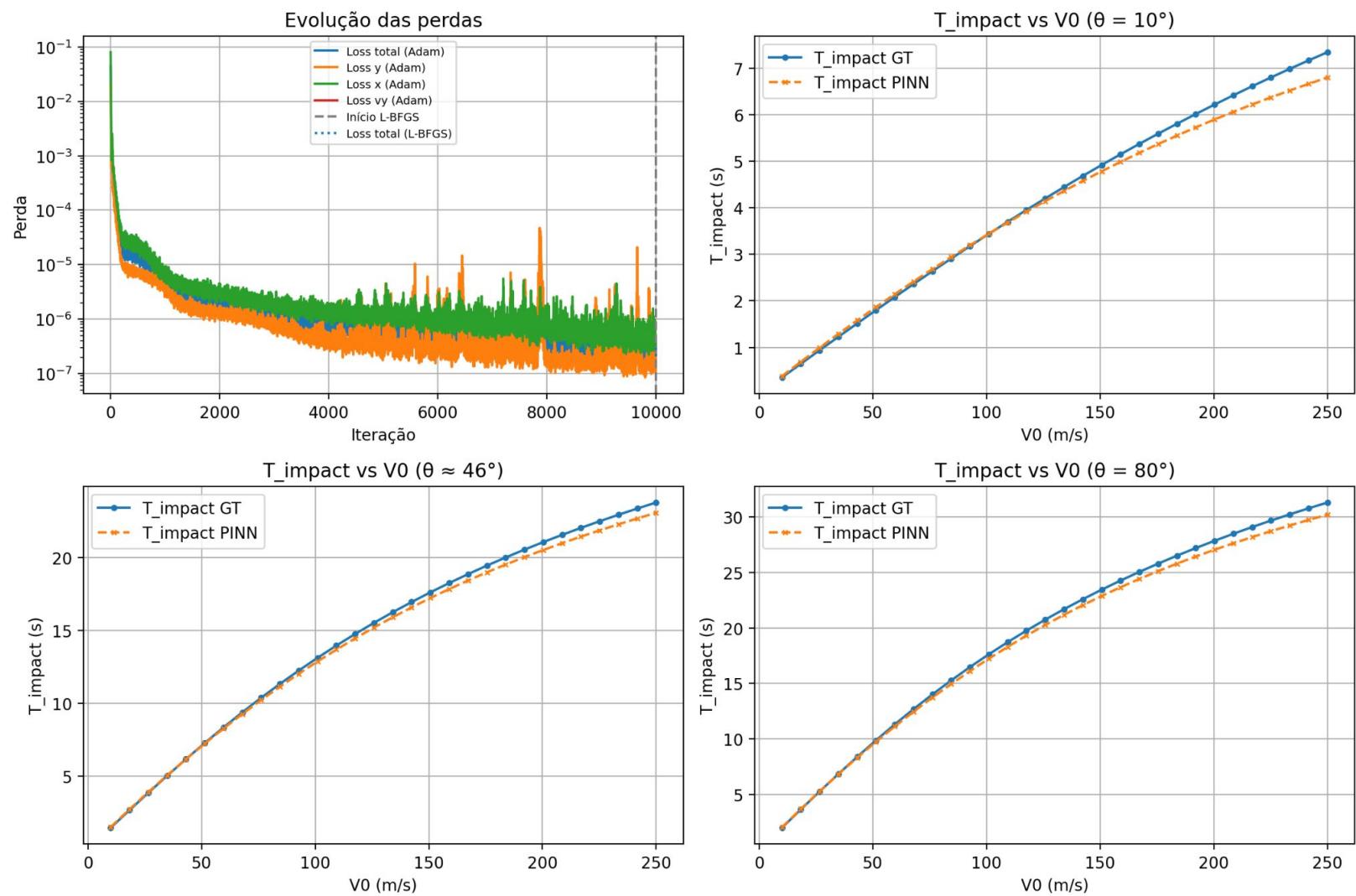
    # Condições iniciais
    vx = v0 * torch.cos(theta0_rad) # (batch,)
    vy = v0 * torch.sin(theta0_rad) # (batch,)
    x = torch.zeros(batch_size, device=device)
    y = torch.zeros(batch_size, device=device)

    # Passo de tempo
    dt = T_impact.squeeze(1) / self.n_integration # (batch,)

    # Integração Euler
    for _ in range(self.n_integration):
        speed = torch.sqrt(vx**2 + vy**2 + 1e-8) # evita divisão por zero
```

Comparação PINN vs Ground Truth (com arrasto, Cd=0.47)  
RMSE T: 0.3998s | RMSE X: 27.3557m





```
=====
Iniciando treinamento com Adam...
```

```
Parâmetros de arrasto: Cd=0.47, k=0.000690
```

```
=====
[Adam] step=20000  loss=3.217063e-08  L_y=3.080e-08  L_x=4.069e-08  L_vy=0.000e+00
[Adam] step=20500  loss=9.668148e-08  L_y=1.380e-07  L_x=7.685e-08  L_vy=0.000e+00
[Adam] step=21000  loss=5.218972e-07  L_y=8.387e-07  L_x=3.211e-07  L_vy=0.000e+00
[Adam] step=21500  loss=1.051180e-07  L_y=1.722e-07  L_x=6.140e-08  L_vy=0.000e+00
[Adam] step=22000  loss=4.874240e-08  L_y=4.686e-08  L_x=6.146e-08  L_vy=0.000e+00
[Adam] step=22500  loss=4.403173e-08  L_y=2.351e-08  L_x=7.434e-08  L_vy=0.000e+00
[Adam] step=23000  loss=1.688441e-07  L_y=2.102e-07  L_x=1.650e-07  L_vy=0.000e+00
[Adam] step=23500  loss=3.462886e-08  L_y=2.861e-08  L_x=4.834e-08  L_vy=0.000e+00
[Adam] step=24000  loss=6.814546e-07  L_y=8.969e-07  L_x=6.174e-07  L_vy=0.000e+00
[Adam] step=24500  loss=2.501101e-07  L_y=5.227e-07  L_x=3.310e-08  L_vy=0.000e+00
```

```
✓ # ===== (Opcional) Refinamento com L-BFGS ===== ...
```

```
=====
Iniciando refinamento com L-BFGS (max 3000 iterações)...
```

```
=====
[L-BFGS] iter=001  loss=1.145920e-07  L_y=1.586e-07  L_x=9.605e-08  L_vy=0.000e+00
L-BFGS finalizado após 8 iterações
Loss final L-BFGS: 2.551184e-08
```

...  
Checkpoint salvo em D:\Code\cannon\_shot\_pinn\checkpoints\impact\_arrasto\impact\_arrasto-2\_64  
Melhor loss: 2.145501e-08 no passo 24822

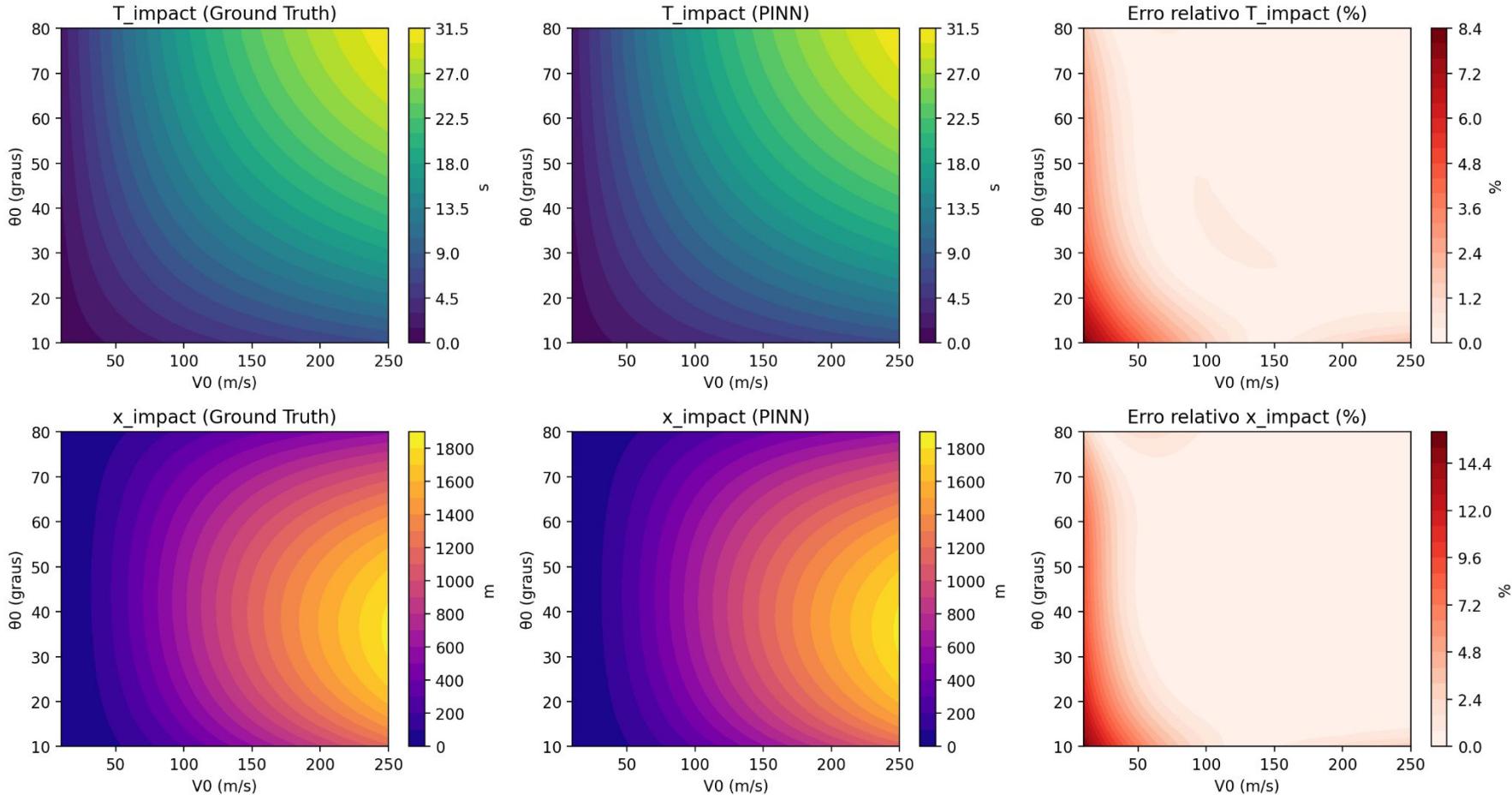
Treino finalizado.

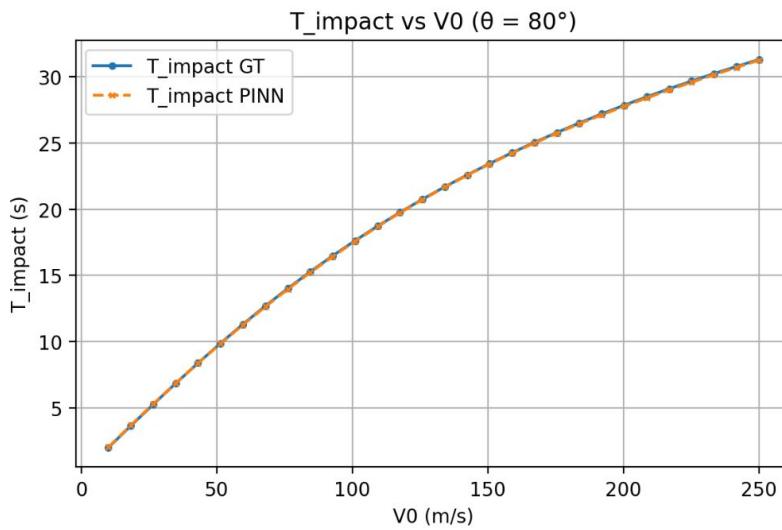
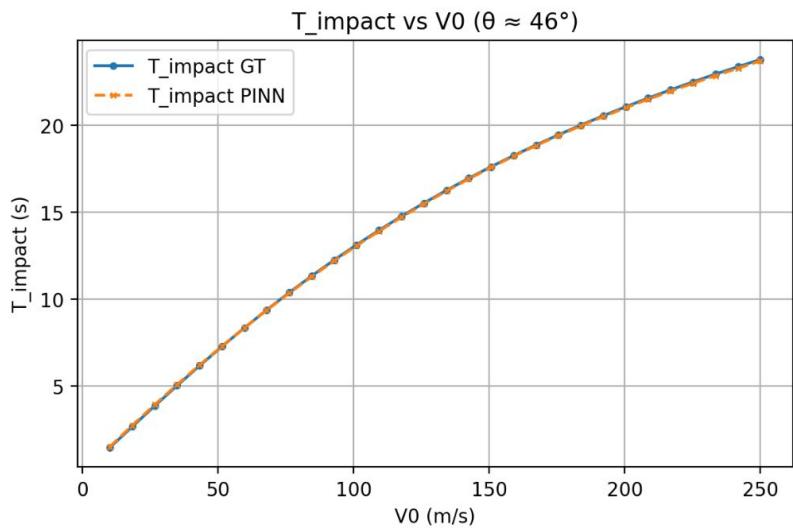
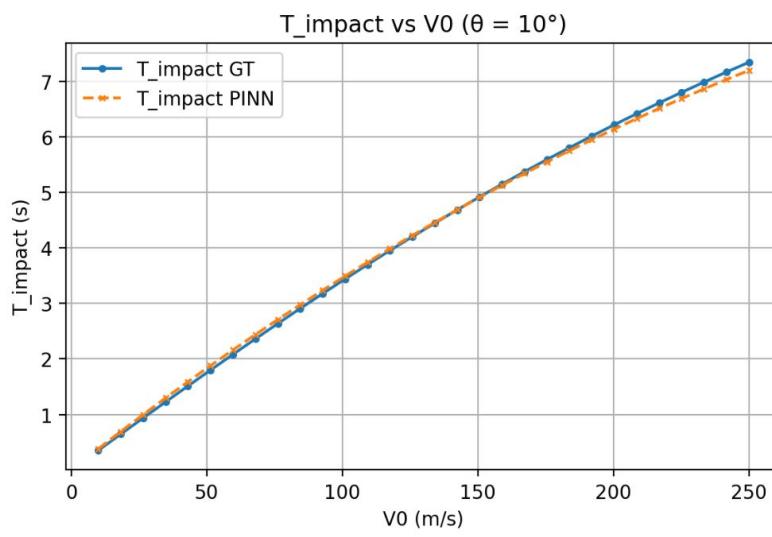
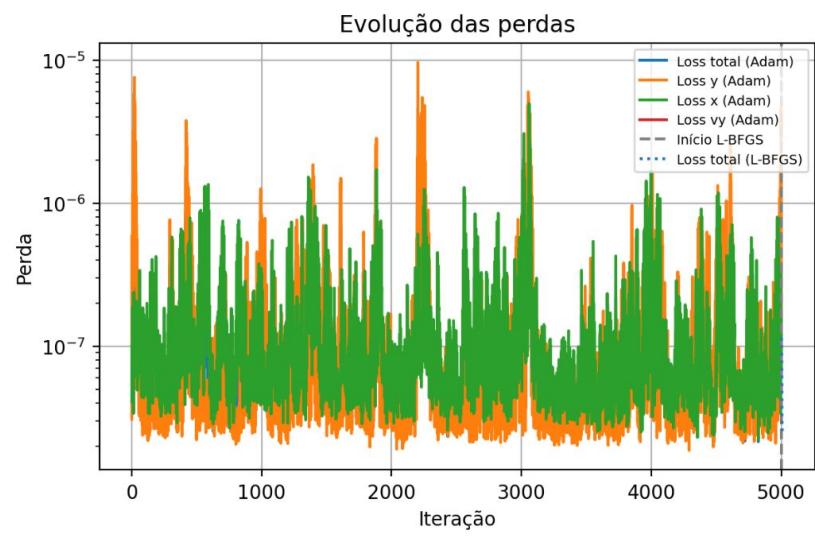
✓ # ===== Usar melhor modelo ===== ...  
... Usando melhor modelo (passo 24822) para avaliação

✓ # ===== Avaliação ===== ...  
...  
Calculando ground truth (pode demorar)...  
Ground truth calculado.

RMSE T\_impact: 5.4946e-02 s  
RMSE x\_impact: 4.0144e+00 m  
Erro relativo médio T: 0.76%  
Erro relativo médio X: 1.23%

Comparação PINN vs Ground Truth (com arrasto, Cd=0.47)  
RMSE T: 0.0549s | RMSE X: 4.0144m





=====

PROBLEMA INVERSO: Dado  $x_{target}$ , encontrar  $(V_0, \theta, T_{impact})$

=====

Objetivo: Minimizar  $V_0$  (com arrasto,  $\theta$  ótimo <  $45^\circ$ )

$x_{target}$	$V_0$	$\theta$	$x_{PINN}$	$x_{real}$	$V_0$ Anal.	$\theta$ Anal.	Status
100	31.70	$44.37^\circ$	100.0	97.1	32.19	$44.6^\circ$	X Real:2.9%
500	81.26	$42.73^\circ$	500.0	503.2	80.92	$42.8^\circ$	✓ OK
1000	135.02	$40.27^\circ$	1000.0	1001.7	134.82	$40.3^\circ$	✓ OK
1400	184.84	$38.13^\circ$	1400.0	1402.6	184.49	$38.1^\circ$	✓ OK
1800	245.72	$36.03^\circ$	1800.0	1804.5	244.96	$36.0^\circ$	✓ OK

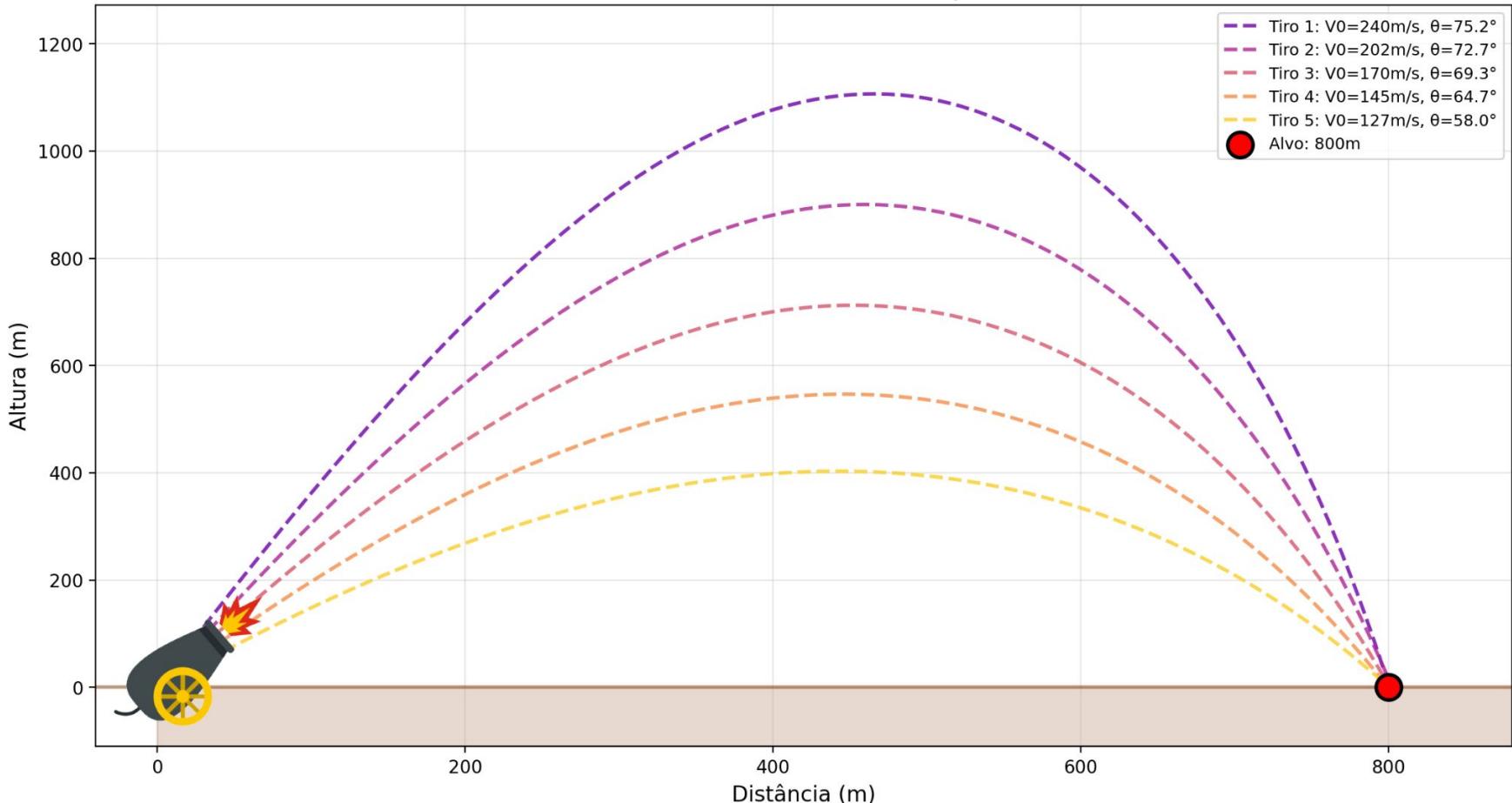
Objetivo: Ângulo fixo de  $54.0^\circ$

$x_{target}$	$V_0$	$\theta$	$x_{PINN}$	$x_{real}$	$V_0$ Anal.	$\theta$ Anal.	Status
100	32.64	$54.00^\circ$	100.0	97.3	33.11	$54.0^\circ$	X Real:2.7%
500	84.93	$54.00^\circ$	500.0	502.0	84.69	$54.0^\circ$	✓ OK
1000	146.33	$54.00^\circ$	1000.0	1000.5	146.26	$54.0^\circ$	✓ OK
1400	210.40	$54.00^\circ$	1400.0	1405.4	209.42	$54.0^\circ$	✓ OK
1800	Não encontrado		---	---	Impossível		

Todas as soluções para x\_target = 800m (PINN vs Ground Truth)

$\theta$ (°)	V0 PINN	T PINN	V0 GT	T GT	Erro V0 %
10.0	187.9	5.85	185.6	5.85	1.23
15.0	153.9	7.21	153.8	7.22	0.08
20.0	136.2	8.42	136.0	8.42	0.16
25.0	125.4	9.52	125.1	9.54	0.25
30.0	118.6	10.60	118.3	10.63	0.26
35.0	114.7	11.70	114.4	11.72	0.28
40.0	113.1	12.84	112.7	12.86	0.30
45.0	113.6	14.06	113.3	14.08	0.29
50.0	116.3	15.41	116.0	15.44	0.22
55.0	121.7	16.96	121.6	17.00	0.10
60.0	130.9	18.83	131.0	18.88	0.02
65.0	146.6	21.23	146.7	21.28	0.05
70.0	175.2	24.57	174.9	24.61	0.15
75.0	236.8	29.80	236.5	29.87	0.15

BATERIA DE TIROS - Impacto Simultâneo em  $x=800\text{m}$  (com arrasto)  
5 tiros com  $\Delta t=3.0\text{s}$  entre disparos



## RESULTADOS

Método	T_impact (s)	x_impact (m)	Tempo (ms)	± σ	Speedup
PINN (rede neural)	17.2115	1113.65	0.7142	0.4129	REF
solve_ivp RK45	17.2523	1115.00	1.6401	0.4085	2.3x
solve_ivp RK23	17.2548	1114.86	2.4854	1.1846	3.5x
solve_ivp DOP853	17.2500	1114.77	2.6300	0.7289	3.7x
Euler (dt=0.001)	17.2490	1114.67	57.5397	15.3974	80.6x
Euler (dt=0.01)	17.2400	1113.73	6.2523	2.5841	8.8x
RK4 manual (dt=0.01)	17.2600	1115.18	68.2141	7.7383	95.5x
RK4 manual (dt=0.001)	17.2510	1114.81	683.8556	56.2365	957.5x

# PRÓXIMOS PASSOS??

Fazer um operador neural

`X_impact => [vθ(θθ), T_Impact(θθ)]`