

TOP_BKN_NCLT_dataset

December 26, 2025

```
[1]: from pathlib import Path
from scipy.io import loadmat
import sys
import os

notebook_path = os.getcwd()
print (f"Current notebook path: {notebook_path}")
project_root = os.path.dirname(notebook_path)
if project_root not in sys.path:
    sys.path.insert(0, project_root)
print (f"Added {project_root} to sys.path")
```

Current notebook path: /home/luky/skola/KalmanNet-main/navigation NCLT dataset
Added /home/luky/skola/KalmanNet-main to sys.path

```
[2]: import torch
import matplotlib.pyplot as plt
from utils import trainer
from utils import utils
from Systems import DynamicSystem
import Filters
import torch.nn.functional as F
from torch.utils.data import TensorDataset, DataLoader
import numpy as np
from scipy.io import loadmat
from scipy.interpolate import RegularGridInterpolator
import random

torch.manual_seed(42)
np.random.seed(42)
random.seed(42)
if torch.cuda.is_available():
    torch.cuda.manual_seed_all(42)

device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
DEVICE = device # For backward compatibility
print(f"device: {device}")
```

```
device: cuda
```

```
[3]: import torch
import torch.nn as nn
import torch.nn.functional as F
from torch.utils.data import TensorDataset, DataLoader
import numpy as np
import matplotlib.pyplot as plt
import os
import Systems

# Parametry sekvencí
TRAIN_SEQ_LEN = 100          # Délka sekvence pro trénink (např. 100 kroků = 100 sekund při 1Hz)
VAL_SEQ_LEN = 250
TEST_SEQ_LEN = 1000           # Délka sekvence pro testování (delší sekvence pro stabilnější vyhodnocení)
STRIDE = 20                   # Posun okna (překryv) pro data augmentation
BATCH_SIZE = 128
DATA_PATH = 'data/processed'
print(f"Běží na zařízení: {device}")
```

Běží na zařízení: cuda

```
[4]: import torch
from torch.utils.data import TensorDataset, DataLoader
import os

def prepare_sequences(dataset_list, seq_len, stride, mode='train'):
    """
    Zpracuje list trajektorií na sekvence pro trénink dle článku.

    Nový formát dle [Song et al., 2024]:
    - Vstup u (4D): [v_left, v_right, theta_imu, omega_imu]
    - Cíl x (6D): [px, py, vx, vy, theta, omega]
    """

    X_seq_list = [] # Ground Truth (Cíl)
    Y_seq_list = [] # GPS Měření (Vstup do korekce)
    U_seq_list = [] # Control Input (IMU/Odo)

    print(f"Zpracovávám {len(dataset_list)} trajektorií pro {mode}...")

    for traj in dataset_list:
        # 1. Extrahuje data
        # GT z preprocessingu je [px, py, theta]
        gt = traj['ground_truth'].float()
```

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# GPS: [x, y] (obsahuje NaN!)
gps = traj['filtered_gps'].float()

# IMU: [ax, ay, theta, omega]
imu = traj['imu'].float()
theta_imu = imu[:, 2] # Orientace z IMU
omega_imu = imu[:, 3] # Úhlová rychlosť z IMU

# ODO: [v_left, v_right]
odo = traj['filtered_wheel'].float()

# Fix NaN v odometrii (nahradíme nulou)
v_left = torch.nan_to_num(odo[:, 0], nan=0.0)
v_right = torch.nan_to_num(odo[:, 1], nan=0.0)

# 2. Sestavení vstupu  $u = [v_l, v_r, \theta_{imu}, \omega_{imu}]$  (4D)
# Toto odpovídá "State Model" definovanému v článku (sekce II.C.2)
u = torch.stack((v_left, v_right, theta_imu, omega_imu), dim=1)

# 3. Sestavení cíle  $x$  (6D) pro state vector [px, py, vx, vy, theta, omega]
# Vyplníme to, co máme z Ground Truth (px, py, theta).
# Rychlosti (vx, vy, omega) v GT implicitně nemáme (nebo je složité jaderivovať priesne),
# ale pro trénink Loss funkce budeme stejně porovnávat primárně pozici.
T = gt.shape[0]
x_target = torch.zeros(T, 6)
x_target[:, 0] = gt[:, 0] # px
x_target[:, 1] = gt[:, 1] # py
x_target[:, 4] = gt[:, 2] # theta
# Ostatní (vx, vy, omega) zústávají 0, protože v Loss funkci budeme maskovať nebo bráť len pozici.

# 4. Sliding Window (Rozsekání na sekvence)
num_samples = gt.shape[0]
current_stride = stride if mode == 'train' else seq_len # U testu bez překryvu

for i in range(0, num_samples - seq_len + 1, current_stride):
    # Cíl: 6D stav
    x_seq = x_target[i : i+seq_len, :]

    # Měření: GPS [px, py]
    y_seq = gps[i : i+seq_len, :]

    # Vstup: 4D control input

```

```

    u_seq = u[i : i+seq_len, :]

    X_seq_list.append(x_seq)
    Y_seq_list.append(y_seq)
    U_seq_list.append(u_seq)

# Stack do tenzorů
X_out = torch.stack(X_seq_list)
Y_out = torch.stack(Y_seq_list)
U_out = torch.stack(U_seq_list)

return X_out, Y_out, U_out

# === NAČTENÍ DAT ===
# Ujistíme se, že cesty a konstanty jsou definované (pokud nejsou, doplňte je u
# nahoře)
# if 'DATA_PATH' not in locals(): DATA_PATH = 'data/processed'
# if 'TRAIN_SEQ_LEN' not in locals(): TRAIN_SEQ_LEN = 100
# if 'VAL_SEQ_LEN' not in locals(): VAL_SEQ_LEN = 200
# if 'TEST_SEQ_LEN' not in locals(): TEST_SEQ_LEN = 500
# if 'STRIDE' not in locals(): STRIDE = 20
# if 'BATCH_SIZE' not in locals(): BATCH_SIZE = 256

train_data_raw = torch.load(os.path.join(DATA_PATH, 'train.pt'))
val_data_raw = torch.load(os.path.join(DATA_PATH, 'val.pt'))
test_data_raw = torch.load(os.path.join(DATA_PATH, 'test.pt'))
# Načtení celého balíku
# train_data_full = torch.load(os.path.join(DATA_PATH, 'train.pt'))

# === RYCHLÝ TEST: OŘÍZNUTÍ DAT ===
# Vezmeme jen prvních 5 trajektorií z 22.
# To radikálně zrychlí jednu epochu a umožní rychle otestovat stabilitu u
# hyperparametrů.
# train_data_raw = train_data_full[:10]

print(f"DEBUG: Pro rychlý test používám jen {len(train_data_raw)} trajektorií.")
# ... zbytek kódu beze změny

# === PŘÍPRAVA SEKVENCÍ ===
print("--- Generuje trénovací data (Paper compatible) ---")
train_X, train_Y, train_U = prepare_sequences(train_data_raw, TRAIN_SEQ_LEN, u
# STRIDE, 'train')

print("\n--- Generuje validační data ---")
val_X, val_Y, val_U = prepare_sequences(val_data_raw, VAL_SEQ_LEN, VAL_SEQ_LEN, u
# 'val')

```

```

print("\n--- Generuji testovací data ---")
test_X, test_Y, test_U = prepare_sequences(test_data_raw, TEST_SEQ_LEN,
                                         TEST_SEQ_LEN, 'test')

# Vytvoření DataLoaderů
train_loader = DataLoader(TensorDataset(train_X, train_Y, train_U),
                           batch_size=BATCH_SIZE, shuffle=True)
val_loader = DataLoader(TensorDataset(val_X, val_Y, val_U),
                           batch_size=BATCH_SIZE, shuffle=False)
test_loader = DataLoader(TensorDataset(test_X, test_Y, test_U),
                           batch_size=BATCH_SIZE, shuffle=False)

print(f"\n Data připravena.")
print(f"Train batches: {len(train_loader)}")
print(f"Shapes -> X: {train_X.shape} (6D State), U: {train_U.shape} (4D Input),"
      f" Y: {train_Y.shape} (2D Meas)")

```

DEBUG: Pro rychlý test používám jen 22 trajektorií.

--- Generuji trénovací data (Paper compatible) ---

Zpracovávám 22 trajektorií pro train...

--- Generuji validační data ---

Zpracovávám 2 trajektorií pro val...

--- Generuji testovací data ---

Zpracovávám 3 trajektorií pro test...

Data připravena.

Train batches: 43

Shapes -> X: torch.Size([5446, 100, 6]) (6D State), U: torch.Size([5446, 100, 4]) (4D Input), Y: torch.Size([5446, 100, 2]) (2D Meas)

[5]: # === INICIALIZACE DYNAMICKÉHO MODELU (System Instance - Paper Version) ===

```

# 1. Parametry systému podle článku [Song et al., 2024]
# State (6D): [px, py, vx, vy, theta, omega]
# Referenční rovnice (5) v článku.
state_dim = 6
# Meas (2D): [gps_x, gps_y]
# Referenční rovnice (6) v článku.
obs_dim = 2
# Časový krok (z preprocessingu)
dt = 1.0

# 2. Definice Matice Q (Procesní šum / Model Uncertainty)
# Nyní máme 6 stavů. Musíme definovat nejistotu pro každý z nich.
# Hodnoty jsou nastaveny heuristicky (lze ladit):

```

```

# - Pozice (idx 0,1): 0.1
# - Rychlost (idx 2,3): 0.1
# - Úhel/Omega (idx 4,5): 0.01 (IMU je v NCLT docela přesné, ale driftuje)
q_diag = torch.tensor([0.1, 0.1, 0.1, 0.1, 0.01, 0.01])
Q = torch.diag(q_diag)

# 3. Definice Matice R (Šum měření / Sensor Noise)
# GPS měří jen pozici (px, py).
# Nastavujeme 1.0 m^2. To odpovídá standardní odchylce 1m.
# Pokud je GPS v datasetu horší, KalmanNet se naučí "nedůvěřovat" vstupu y
# a spoléhat více na predikci z u (odometrie).
r_diag = torch.tensor([1.0, 1.0])
R = torch.diag(r_diag)

# 4. Počáteční podmínky (Prior)
# Ex0: Nulový vektor 6x1
Ex0 = torch.zeros(state_dim, 1)

# P0: Počáteční kovariance
# Autoři používají P k inicializaci EKF[cite: 700].
# Nastavíme rozumnou počáteční nejistotu.
P0 = torch.eye(state_dim) * 0.5

# 5. Vytvoření instance DynamicSystemNCLT
# Důležité: f=None zajistí, že se použije interní `_f_paper_dynamics` (rovnice ↵5),
# která očekává 4D vstup (v_l, v_r, theta, omega).
sys_model = Systems.DynamicSystemNCLT(
    state_dim=state_dim,
    obs_dim=obs_dim,
    Q=Q,
    R=R,
    Ex0=Ex0,
    P0=P0,
    dt=dt,
    f=None, # None -> Použije se model z článku: px += vc*cos(theta_imu)...
    h=None, # None -> Použije se GPS model: y = [px, py]
    device=DEVICE
)

print(f" System Model NCLT inicializován (Paper Version).")
print(f" - State Dim: {sys_model.state_dim} [px, py, vx, vy, theta, omega]")
print(f" - Meas Dim: {sys_model.obs_dim} [gps_x, gps_y]")
print(f" - Input Dim: 4 [v_l, v_r, theta_imu, omega_imu]") # Implicitně v modelu
print(f" - Q Diag: {q_diag.tolist()}")

```

System Model NCLT inicializován (Paper Version).

- State Dim: 6 [px, py, vx, vy, theta, omega]
- Meas Dim: 2 [gps_x, gps_y]
- Input Dim: 4 [v_l, v_r, theta_imu, omega_imu]
- Q Diag: [0.10000000149011612, 0.10000000149011612, 0.10000000149011612, 0.10000000149011612, 0.009999999776482582, 0.009999999776482582]

```
[7]: import torch
import torch.optim as optim
import os
from state_NN_models import NCLT
from utils import trainer

state_knet = NCLT.StateBayesianKalmanNetNCLT(
    system_model=sys_model,
    device=DEVICE,
    hidden_size_multiplier=8,           # Větší kapacita pro složitější dynamiku
    output_layer_multiplier=4,
    num_gru_layers=1,                  # 1 vrstva GRU obvykle stačí a je u
    ↪stabilnější
    init_min_dropout=0.2,
    init_max_dropout=0.4
).to(DEVICE)
print(state_knet)

# Počet trénovatelných parametrů
params_count = sum(p.numel() for p in state_knet.parameters() if p.
    ↪requires_grad)
print(f"Model má {params_count} trénovatelných parametrů.")
# === 3. PRÍPRAVA A SPUŠTĚNÍ TRÉNINKU ===
print("\n Spouštím tréninkovou smyčku (Vektorizovanou)...")

# 1. Zjistíme velikost datasetu (počet batchů v jedné epoše)
num_train_batches = len(train_loader)
print(f"INFO: 1 Epocha = {num_train_batches} iterací (batchů).")

# 2. Nastavení parametrů
EPOCHS = 100
WARMUP_EPOCHS = 15 # Prvních 15 epoch jen MSE, aby se model stabilizoval

# Přepočet na iterace (protože nová funkce počítá v iteracích)
total_iterations = EPOCHS * num_train_batches
total_iterations = 800
warmup_iterations = WARMUP_EPOCHS * num_train_batches
warmup_iterations=350
validation_period = 5 # Chceme validovat jednou za epochu (na konci)

# 3. Volání funkce
```

```

trained_knet = trainer.train_BayesianKalmanNet(
    model=state_knet,
    train_loader=train_loader,
    val_loader=val_loader,
    device=DEVICE,

    # Parametry iterací
    total_train_iter=total_iterations,
    learning_rate=5e-4,           # Opatrný LR pro BKN
    clip_grad=1.0,                # Stabilizace gradientů
    J_samples=10,                  # Počet MC vzorků (paralelně)

    # Logování a validace
    validation_period=validation_period,
    logging_period=10,            # Výpis loss každých 10 batchů

    # Warmup
    warmup_iterations=warmup_iterations,
    weight_decay_=1e-5
)

```

INFO: Aplikuji 'Start Zero' inicializaci pro Kalman Gain.
 DEBUG: Výstupní vrstva vynulována (Soft Start).
 StateBayesianKalmanNetNCLT(
 (dnn): DNN_BayesianKalmanNetNCLT(
 (input_layer): Sequential(
 (0): Linear(in_features=16, out_features=512, bias=True)
 (1): ReLU()
)
 (concrete_dropout1): ConcreteDropout()
 (gru): GRU(512, 160)
 (output_layer): Sequential(
 (0): Linear(in_features=160, out_features=48, bias=True)
 (1): ReLU()
 (2): Linear(in_features=48, out_features=12, bias=True)
)
 (concrete_dropout2): ConcreteDropout()
)
)
 Model má 340542 trénovatelných parametrů.

Spouštím tréninkovou smyčku (Vektorizovanou)...
 INFO: 1 Epocha = 43 iterací (batchů).

[DIAGNOSTIKA iter 0]
 -> Rozdíl (Pred - Target): 1.813148
 -> JE TO PŘESNÁ NULA? Ne

```
-> Min Variance v batchi: 2.47e-03

--- Validation at iteration 5 ---
Average MSE: 20.8313, Average ANEES: 335.7215
>>> New best VALIDATION ANEES! Saving model. <<<
-----

--- Iteration [10/800] ---
- Total Loss: 7.7659
- NLL: 0.0000
- Reg: 0.1679
- p1=0.374, p2=0.399

--- Validation at iteration 10 ---
Average MSE: 37.3800, Average ANEES: 191.9882
>>> New best VALIDATION ANEES! Saving model. <<<
-----

[DIAGNOSTIKA iter 10]
-> Rozdíl (Pred - Target): 1.335995
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.74e-03

--- Validation at iteration 15 ---
Average MSE: 21.0435, Average ANEES: 246.1053
-----

--- Iteration [20/800] ---
- Total Loss: 7.6589
- NLL: 0.0000
- Reg: 0.1675
- p1=0.374, p2=0.398

--- Validation at iteration 20 ---
Average MSE: 14.5628, Average ANEES: 227.1221
-----

[DIAGNOSTIKA iter 20]
-> Rozdíl (Pred - Target): 1.152306
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 3.25e-03

--- Validation at iteration 25 ---
Average MSE: 8.2143, Average ANEES: 188.3475
>>> New best VALIDATION ANEES! Saving model. <<<
-----

--- Iteration [30/800] ---
- Total Loss: 4.5368
- NLL: 0.0000
- Reg: 0.1671
```

```
- p1=0.373, p2=0.397

--- Validation at iteration 30 ---
Average MSE: 15.4657, Average ANEES: 263.3331
-----

[DIAGNOSTIKA iter 30]
-> Rozdíl (Pred - Target): 1.159997
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.43e-03

--- Validation at iteration 35 ---
Average MSE: 12.8131, Average ANEES: 198.9188
-----

--- Iteration [40/800] ---
- Total Loss: 6.9037
- NLL: 0.0000
- Reg: 0.1666
- p1=0.372, p2=0.396

--- Validation at iteration 40 ---
Average MSE: 15.6414, Average ANEES: 322.0733
-----

[DIAGNOSTIKA iter 40]
-> Rozdíl (Pred - Target): 1.199098
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.14e-03

--- Validation at iteration 45 ---
Average MSE: 11.3383, Average ANEES: 185.9488
>>> New best VALIDATION ANEES! Saving model. <<<
-----

--- Iteration [50/800] ---
- Total Loss: 5.4543
- NLL: 0.0000
- Reg: 0.1661
- p1=0.371, p2=0.396

--- Validation at iteration 50 ---
Average MSE: 11.5102, Average ANEES: 218.3388
-----

[DIAGNOSTIKA iter 50]
-> Rozdíl (Pred - Target): 1.112368
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 1.02e-03
```

```
--- Validation at iteration 55 ---  
Average MSE: 10.1489, Average ANEES: 159.8814  
>>> New best VALIDATION ANEES! Saving model. <<<
```

```
-----  
--- Iteration [60/800] ---  
- Total Loss: 6.2605  
- NLL: 0.0000  
- Reg: 0.1656  
- p1=0.370, p2=0.395
```

```
--- Validation at iteration 60 ---  
Average MSE: 12.2046, Average ANEES: 124.0069  
>>> New best VALIDATION ANEES! Saving model. <<<
```

```
[DIAGNOSTIKA iter 60]  
-> Rozdíl (Pred - Target): 1.193260  
-> JE TO PŘESNÁ NULA? Ne  
-> Min Variance v batchi: 3.54e-03
```

```
--- Validation at iteration 65 ---  
Average MSE: 8.3135, Average ANEES: 149.7909
```

```
-----  
--- Iteration [70/800] ---  
- Total Loss: 4.8238  
- NLL: 0.0000  
- Reg: 0.1652  
- p1=0.369, p2=0.394
```

```
--- Validation at iteration 70 ---  
Average MSE: 11.2442, Average ANEES: 122.7932  
>>> New best VALIDATION ANEES! Saving model. <<<
```

```
[DIAGNOSTIKA iter 70]  
-> Rozdíl (Pred - Target): 1.024722  
-> JE TO PŘESNÁ NULA? Ne  
-> Min Variance v batchi: 3.13e-03
```

```
--- Validation at iteration 75 ---  
Average MSE: 11.9845, Average ANEES: 157.0967
```

```
-----  
--- Iteration [80/800] ---  
- Total Loss: 4.4905  
- NLL: 0.0000  
- Reg: 0.1649  
- p1=0.368, p2=0.393
```

```
--- Validation at iteration 80 ---  
Average MSE: 12.8258, Average ANEES: 135.1505
```

```
[DIAGNOSTIKA iter 80]  
-> Rozdíl (Pred - Target): 0.998969  
-> JE TO PŘESNÁ NULA? Ne  
-> Min Variance v batchi: 4.13e-03
```

```
--- Validation at iteration 85 ---  
Average MSE: 13.7881, Average ANEES: 132.0691
```

```
--- Iteration [90/800] ---  
- Total Loss: 4.8382  
- NLL: 0.0000  
- Reg: 0.1648  
- p1=0.368, p2=0.393
```

```
--- Validation at iteration 90 ---  
Average MSE: 10.8060, Average ANEES: 140.2592
```

```
[DIAGNOSTIKA iter 90]  
-> Rozdíl (Pred - Target): 1.015304  
-> JE TO PŘESNÁ NULA? Ne  
-> Min Variance v batchi: 1.80e-03
```

```
--- Validation at iteration 95 ---  
Average MSE: 7.9230, Average ANEES: 137.3890
```

```
--- Iteration [100/800] ---  
- Total Loss: 4.6629  
- NLL: 0.0000  
- Reg: 0.1646  
- p1=0.368, p2=0.393
```

```
--- Validation at iteration 100 ---  
Average MSE: 17.5678, Average ANEES: 115.2179  
>>> New best VALIDATION ANEES! Saving model. <<<
```

```
[DIAGNOSTIKA iter 100]  
-> Rozdíl (Pred - Target): 1.038241  
-> JE TO PŘESNÁ NULA? Ne  
-> Min Variance v batchi: 2.79e-03
```

```
--- Validation at iteration 105 ---  
Average MSE: 10.4049, Average ANEES: 111.3852
```

```

>>> New best VALIDATION ANEES! Saving model. <<<
-----
--- Iteration [110/800] ---
- Total Loss: 4.1243
- NLL: 0.0000
- Reg: 0.1645
- p1=0.369, p2=0.392

--- Validation at iteration 110 ---
Average MSE: 8.2017, Average ANEES: 114.7936
-----

[DIAGNOSTIKA iter 110]
-> Rozdíl (Pred - Target): 0.933707
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 3.31e-03

--- Validation at iteration 115 ---
Average MSE: 10.4019, Average ANEES: 131.8068
-----

--- Iteration [120/800] ---
- Total Loss: 4.4941
- NLL: 0.0000
- Reg: 0.1643
- p1=0.369, p2=0.391

--- Validation at iteration 120 ---
Average MSE: 11.2274, Average ANEES: 118.3047
-----

[DIAGNOSTIKA iter 120]
-> Rozdíl (Pred - Target): 1.018751
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 3.32e-03

--- Validation at iteration 125 ---
Average MSE: 8.4730, Average ANEES: 156.6305
-----

--- Iteration [130/800] ---
- Total Loss: 6.8227
- NLL: 0.0000
- Reg: 0.1641
- p1=0.368, p2=0.390

--- Validation at iteration 130 ---
Average MSE: 13.7900, Average ANEES: 103.3026
>>> New best VALIDATION ANEES! Saving model. <<<
-----
```

```
[DIAGNOSTIKA iter 130]
-> Rozdíl (Pred - Target): 1.030146
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.43e-03

--- Validation at iteration 135 ---
Average MSE: 16.5233, Average ANEES: 115.2235
-----
--- Iteration [140/800] ---
- Total Loss: 4.6046
- NLL: 0.0000
- Reg: 0.1637
- p1=0.367, p2=0.390

--- Validation at iteration 140 ---
Average MSE: 11.0963, Average ANEES: 147.9578
-----

[DIAGNOSTIKA iter 140]
-> Rozdíl (Pred - Target): 1.009480
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 3.11e-03

--- Validation at iteration 145 ---
Average MSE: 7.3468, Average ANEES: 96.2440
>>> New best VALIDATION ANEES! Saving model. <<<
-----
--- Iteration [150/800] ---
- Total Loss: 4.3097
- NLL: 0.0000
- Reg: 0.1634
- p1=0.367, p2=0.389

--- Validation at iteration 150 ---
Average MSE: 10.5350, Average ANEES: 101.3948
-----

[DIAGNOSTIKA iter 150]
-> Rozdíl (Pred - Target): 1.015473
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.61e-03

--- Validation at iteration 155 ---
Average MSE: 10.0813, Average ANEES: 171.5844
-----
--- Iteration [160/800] ---
- Total Loss: 4.2172
```

- NLL: 0.0000
- Reg: 0.1631
- p1=0.366, p2=0.388

--- Validation at iteration 160 ---
Average MSE: 8.4781, Average ANEES: 113.9435

[DIAGNOSTIKA iter 160]
-> Rozdíl (Pred - Target): 0.998008
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 3.25e-03

--- Validation at iteration 165 ---
Average MSE: 9.7962, Average ANEES: 125.4422

--- Iteration [170/800] ---
- Total Loss: 3.9152
- NLL: 0.0000
- Reg: 0.1630
- p1=0.366, p2=0.388

--- Validation at iteration 170 ---
Average MSE: 10.6910, Average ANEES: 124.6607

[DIAGNOSTIKA iter 170]
-> Rozdíl (Pred - Target): 1.055560
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 3.29e-03

--- Validation at iteration 175 ---
Average MSE: 8.3617, Average ANEES: 113.3452

--- Iteration [180/800] ---
- Total Loss: 4.8322
- NLL: 0.0000
- Reg: 0.1628
- p1=0.366, p2=0.387

--- Validation at iteration 180 ---
Average MSE: 6.4444, Average ANEES: 120.8211

[DIAGNOSTIKA iter 180]
-> Rozdíl (Pred - Target): 0.956754
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.00e-03

```
--- Validation at iteration 185 ---
Average MSE: 8.7022, Average ANEES: 101.9891
```

```
-----  
--- Iteration [190/800] ---
- Total Loss: 4.0546
- NLL: 0.0000
- Reg: 0.1626
- p1=0.366, p2=0.387
```

```
--- Validation at iteration 190 ---
Average MSE: 8.1826, Average ANEES: 107.2664
```

```
-----  
[DIAGNOSTIKA iter 190]
-> Rozdíl (Pred - Target): 1.048492
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.85e-03
```

```
--- Validation at iteration 195 ---
Average MSE: 8.3672, Average ANEES: 98.5544
```

```
-----  
--- Iteration [200/800] ---
- Total Loss: 4.9139
- NLL: 0.0000
- Reg: 0.1624
- p1=0.366, p2=0.386
```

```
--- Validation at iteration 200 ---
Average MSE: 6.6355, Average ANEES: 100.7216
```

```
-----  
[DIAGNOSTIKA iter 200]
-> Rozdíl (Pred - Target): 0.912643
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.90e-03
```

```
--- Validation at iteration 205 ---
Average MSE: 8.0258, Average ANEES: 108.0961
```

```
-----  
--- Iteration [210/800] ---
- Total Loss: 3.5279
- NLL: 0.0000
- Reg: 0.1622
- p1=0.365, p2=0.386
```

```
--- Validation at iteration 210 ---
Average MSE: 6.5842, Average ANEES: 103.3507
```

```
-----  
[DIAGNOSTIKA iter 210]  
-> Rozdíl (Pred - Target): 1.082328  
-> JE TO PŘESNÁ NULA? Ne  
-> Min Variance v batchi: 2.89e-03  
  
--- Validation at iteration 215 ---  
Average MSE: 9.6016, Average ANEES: 111.7953  
-----  
--- Iteration [220/800] ---  
- Total Loss: 3.3752  
- NLL: 0.0000  
- Reg: 0.1619  
- p1=0.365, p2=0.385  
  
--- Validation at iteration 220 ---  
Average MSE: 10.0815, Average ANEES: 81.9325  
>>> New best VALIDATION ANEES! Saving model. <<<  
-----  
[DIAGNOSTIKA iter 220]  
-> Rozdíl (Pred - Target): 1.016657  
-> JE TO PŘESNÁ NULA? Ne  
-> Min Variance v batchi: 1.84e-03  
  
--- Validation at iteration 225 ---  
Average MSE: 11.5836, Average ANEES: 101.2669  
-----  
--- Iteration [230/800] ---  
- Total Loss: 4.1757  
- NLL: 0.0000  
- Reg: 0.1616  
- p1=0.364, p2=0.384  
  
--- Validation at iteration 230 ---  
Average MSE: 10.6988, Average ANEES: 116.2455  
-----  
[DIAGNOSTIKA iter 230]  
-> Rozdíl (Pred - Target): 1.063309  
-> JE TO PŘESNÁ NULA? Ne  
-> Min Variance v batchi: 1.24e-03  
  
--- Validation at iteration 235 ---  
Average MSE: 11.7545, Average ANEES: 89.1696  
-----  
--- Iteration [240/800] ---
```

```
- Total Loss: 4.3265
- NLL: 0.0000
- Reg: 0.1615
- p1=0.364, p2=0.384

--- Validation at iteration 240 ---
Average MSE: 8.9114, Average ANEES: 107.0952
-----

[DIAGNOSTIKA iter 240]
-> Rozdíl (Pred - Target): 0.972190
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 1.42e-03

--- Validation at iteration 245 ---
Average MSE: 6.5866, Average ANEES: 98.5746
-----

--- Iteration [250/800] ---
- Total Loss: 5.6336
- NLL: 0.0000
- Reg: 0.1613
- p1=0.364, p2=0.383

--- Validation at iteration 250 ---
Average MSE: 7.0785, Average ANEES: 108.4219
-----

[DIAGNOSTIKA iter 250]
-> Rozdíl (Pred - Target): 1.041114
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.99e-03

--- Validation at iteration 255 ---
Average MSE: 8.5588, Average ANEES: 78.0972
>>> New best VALIDATION ANEES! Saving model. <<<
-----

--- Iteration [260/800] ---
- Total Loss: 3.9046
- NLL: 0.0000
- Reg: 0.1612
- p1=0.364, p2=0.383

--- Validation at iteration 260 ---
Average MSE: 8.0225, Average ANEES: 71.2592
>>> New best VALIDATION ANEES! Saving model. <<<
-----

[DIAGNOSTIKA iter 260]
```

```
-> Rozdíl (Pred - Target): 1.092668
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.81e-03

--- Validation at iteration 265 ---
Average MSE: 10.8454, Average ANEES: 164.0609
-----

--- Iteration [270/800] ---
- Total Loss: 12.4752
- NLL: 0.0000
- Reg: 0.1610
- p1=0.364, p2=0.383

--- Validation at iteration 270 ---
Average MSE: 7.3039, Average ANEES: 100.9889
-----

[DIAGNOSTIKA iter 270]
-> Rozdíl (Pred - Target): 0.985548
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 1.41e-03

--- Validation at iteration 275 ---
Average MSE: 13.2476, Average ANEES: 84.0938
-----

--- Iteration [280/800] ---
- Total Loss: 4.3233
- NLL: 0.0000
- Reg: 0.1608
- p1=0.363, p2=0.382

--- Validation at iteration 280 ---
Average MSE: 11.1309, Average ANEES: 84.9196
-----

[DIAGNOSTIKA iter 280]
-> Rozdíl (Pred - Target): 1.060897
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 3.62e-03

--- Validation at iteration 285 ---
Average MSE: 8.7348, Average ANEES: 105.7923
-----

--- Iteration [290/800] ---
- Total Loss: 5.4456
- NLL: 0.0000
- Reg: 0.1607
- p1=0.363, p2=0.382
```

```
--- Validation at iteration 290 ---
Average MSE: 14.7797, Average ANEES: 78.5443
```

```
[DIAGNOSTIKA iter 290]
-> Rozdíl (Pred - Target): 0.930990
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.22e-03
```

```
--- Validation at iteration 295 ---
Average MSE: 7.7121, Average ANEES: 92.5174
```

```
--- Iteration [300/800] ---
- Total Loss: 4.4772
- NLL: 0.0000
- Reg: 0.1606
- p1=0.363, p2=0.381
```

```
--- Validation at iteration 300 ---
Average MSE: 10.3512, Average ANEES: 87.4664
```

```
[DIAGNOSTIKA iter 300]
-> Rozdíl (Pred - Target): 0.946608
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 4.22e-03
```

```
--- Validation at iteration 305 ---
Average MSE: 9.1985, Average ANEES: 109.1824
```

```
--- Iteration [310/800] ---
- Total Loss: 3.8570
- NLL: 0.0000
- Reg: 0.1605
- p1=0.363, p2=0.381
```

```
--- Validation at iteration 310 ---
Average MSE: 12.2542, Average ANEES: 107.7352
```

```
[DIAGNOSTIKA iter 310]
-> Rozdíl (Pred - Target): 0.963159
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 8.40e-04
```

```
--- Validation at iteration 315 ---
Average MSE: 8.9155, Average ANEES: 69.9328
```

```
>>> New best VALIDATION ANEES! Saving model. <<<
-----
--- Iteration [320/800] ---
- Total Loss: 4.1374
- NLL: 0.0000
- Reg: 0.1603
- p1=0.363, p2=0.381

--- Validation at iteration 320 ---
Average MSE: 8.1951, Average ANEES: 98.9796
-----

[DIAGNOSTIKA iter 320]
-> Rozdíl (Pred - Target): 1.064174
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 1.89e-03

--- Validation at iteration 325 ---
Average MSE: 7.7121, Average ANEES: 107.3892
-----

--- Iteration [330/800] ---
- Total Loss: 5.5166
- NLL: 0.0000
- Reg: 0.1602
- p1=0.363, p2=0.380

--- Validation at iteration 330 ---
Average MSE: 6.9812, Average ANEES: 76.3379
-----

[DIAGNOSTIKA iter 330]
-> Rozdíl (Pred - Target): 0.972974
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.80e-03

--- Validation at iteration 335 ---
Average MSE: 7.5561, Average ANEES: 86.1223
-----

--- Iteration [340/800] ---
- Total Loss: 3.6423
- NLL: 0.0000
- Reg: 0.1600
- p1=0.363, p2=0.380

--- Validation at iteration 340 ---
Average MSE: 9.0256, Average ANEES: 72.1655
-----
```

```

[DIAGNOSTIKA iter 340]
-> Rozdíl (Pred - Target): 0.928191
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.81e-03

--- Validation at iteration 345 ---
Average MSE: 7.5566, Average ANEES: 113.3990
-----
--- Iteration [350/800] ---
- Total Loss: 3.1682
- NLL: 0.0000
- Reg: 0.1599
- p1=0.362, p2=0.380

--- Validation at iteration 350 ---
Average MSE: 9.3154, Average ANEES: 61.4551
>>> New best VALIDATION ANEES! Saving model. <<<
-----

[DIAGNOSTIKA iter 350]
-> Rozdíl (Pred - Target): 0.977969
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.92e-03

--- Validation at iteration 355 ---
Average MSE: 16.7465, Average ANEES: 51.6080
>>> New best VALIDATION ANEES! Saving model. <<<
-----
--- Iteration [360/800] ---
- Total Loss: 0.7729
- NLL: 0.6135
- Reg: 0.1594
- p1=0.362, p2=0.380

--- Validation at iteration 360 ---
Average MSE: 8.8914, Average ANEES: 52.3262
-----

[DIAGNOSTIKA iter 360]
-> Rozdíl (Pred - Target): 1.039996
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 3.43e-03

--- Validation at iteration 365 ---
Average MSE: 9.4874, Average ANEES: 62.6277
-----
--- Iteration [370/800] ---
- Total Loss: 0.6034

```

```

- NLL: 0.4451
- Reg: 0.1583
- p1=0.361, p2=0.380

--- Validation at iteration 370 ---
Average MSE: 7.4180, Average ANEES: 47.6930
>>> New best VALIDATION ANEES! Saving model. <<<
-----

[DIAGNOSTIKA iter 370]
-> Rozdíl (Pred - Target): 0.961327
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.54e-03

--- Validation at iteration 375 ---
Average MSE: 9.6993, Average ANEES: 40.5449
>>> New best VALIDATION ANEES! Saving model. <<<
-----

--- Iteration [380/800] ---
- Total Loss: 0.6051
- NLL: 0.4481
- Reg: 0.1570
- p1=0.360, p2=0.379

--- Validation at iteration 380 ---
Average MSE: 8.3917, Average ANEES: 43.6165
-----

[DIAGNOSTIKA iter 380]
-> Rozdíl (Pred - Target): 1.027276
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.94e-03

--- Validation at iteration 385 ---
Average MSE: 7.5494, Average ANEES: 43.5987
-----

--- Iteration [390/800] ---
- Total Loss: 0.6351
- NLL: 0.4797
- Reg: 0.1554
- p1=0.359, p2=0.377

--- Validation at iteration 390 ---
Average MSE: 8.8469, Average ANEES: 37.9530
>>> New best VALIDATION ANEES! Saving model. <<<
-----

[DIAGNOSTIKA iter 390]

```

```
-> Rozdíl (Pred - Target): 0.989167
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi:  2.87e-03

--- Validation at iteration 395 ---
Average MSE: 9.1170, Average ANEES: 43.7458
-----

--- Iteration [400/800] ---
- Total Loss: 0.5759
- NLL: 0.4219
- Reg: 0.1540
- p1=0.357, p2=0.376

--- Validation at iteration 400 ---
Average MSE: 8.9849, Average ANEES: 47.0975
-----

[DIAGNOSTIKA iter 400]
-> Rozdíl (Pred - Target): 0.962401
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi:  2.03e-03

--- Validation at iteration 405 ---
Average MSE: 7.8931, Average ANEES: 39.6240
-----

--- Iteration [410/800] ---
- Total Loss: 0.5657
- NLL: 0.4133
- Reg: 0.1524
- p1=0.356, p2=0.374

--- Validation at iteration 410 ---
Average MSE: 7.5488, Average ANEES: 42.5080
-----

[DIAGNOSTIKA iter 410]
-> Rozdíl (Pred - Target): 0.987789
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi:  3.18e-03

--- Validation at iteration 415 ---
Average MSE: 9.3636, Average ANEES: 43.3083
-----

--- Iteration [420/800] ---
- Total Loss: 0.4415
- NLL: 0.2907
- Reg: 0.1508
- p1=0.354, p2=0.372
```

```
--- Validation at iteration 420 ---
Average MSE: 11.9026, Average ANEES: 45.0408
```

```
[DIAGNOSTIKA iter 420]
-> Rozdíl (Pred - Target): 0.882217
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 1.80e-03
```

```
--- Validation at iteration 425 ---
Average MSE: 11.5523, Average ANEES: 54.3717
```

```
--- Iteration [430/800] ---
- Total Loss: 0.4880
- NLL: 0.3386
- Reg: 0.1494
- p1=0.353, p2=0.371
```

```
--- Validation at iteration 430 ---
Average MSE: 9.9588, Average ANEES: 46.9922
```

```
[DIAGNOSTIKA iter 430]
-> Rozdíl (Pred - Target): 0.923186
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.43e-03
```

```
--- Validation at iteration 435 ---
Average MSE: 10.6370, Average ANEES: 39.9162
```

```
--- Iteration [440/800] ---
- Total Loss: 0.5576
- NLL: 0.4099
- Reg: 0.1477
- p1=0.352, p2=0.369
```

```
--- Validation at iteration 440 ---
Average MSE: 10.4837, Average ANEES: 42.7587
```

```
[DIAGNOSTIKA iter 440]
-> Rozdíl (Pred - Target): 1.008402
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.00e-03
```

```
--- Validation at iteration 445 ---
Average MSE: 9.4744, Average ANEES: 42.3872
```

```
-----  
--- Iteration [450/800] ---  
- Total Loss: 0.5388  
- NLL: 0.3928  
- Reg: 0.1461  
- p1=0.351, p2=0.367  
  
--- Validation at iteration 450 ---  
Average MSE: 9.2553, Average ANEES: 43.3609  
-----
```

```
[DIAGNOSTIKA iter 450]  
-> Rozdíl (Pred - Target): 0.998328  
-> JE TO PŘESNÁ NULA? Ne  
-> Min Variance v batchi: 2.63e-03
```

```
--- Validation at iteration 455 ---  
Average MSE: 8.3684, Average ANEES: 46.6202  
-----
```

```
--- Iteration [460/800] ---  
- Total Loss: 0.5243  
- NLL: 0.3798  
- Reg: 0.1445  
- p1=0.349, p2=0.366
```

```
--- Validation at iteration 460 ---  
Average MSE: 10.9624, Average ANEES: 46.9817  
-----
```

```
[DIAGNOSTIKA iter 460]  
-> Rozdíl (Pred - Target): 0.925160  
-> JE TO PŘESNÁ NULA? Ne  
-> Min Variance v batchi: 1.84e-03
```

```
--- Validation at iteration 465 ---  
Average MSE: 10.1335, Average ANEES: 46.5213  
-----
```

```
--- Iteration [470/800] ---  
- Total Loss: 0.3866  
- NLL: 0.2435  
- Reg: 0.1431  
- p1=0.348, p2=0.364
```

```
--- Validation at iteration 470 ---  
Average MSE: 10.4531, Average ANEES: 41.1133  
-----
```

```
[DIAGNOSTIKA iter 470]
```

```
-> Rozdíl (Pred - Target): 0.873532
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi:  1.69e-03

--- Validation at iteration 475 ---
Average MSE: 10.0633, Average ANEES: 48.5902
-----

--- Iteration [480/800] ---
- Total Loss: 0.4921
- NLL: 0.3504
- Reg: 0.1417
- p1=0.347, p2=0.363

--- Validation at iteration 480 ---
Average MSE: 9.9389, Average ANEES: 38.7740
-----

[DIAGNOSTIKA iter 480]
-> Rozdíl (Pred - Target): 0.891287
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi:  3.14e-03

--- Validation at iteration 485 ---
Average MSE: 8.1234, Average ANEES: 40.8472
-----

--- Iteration [490/800] ---
- Total Loss: 0.4610
- NLL: 0.3206
- Reg: 0.1404
- p1=0.346, p2=0.361

--- Validation at iteration 490 ---
Average MSE: 7.7446, Average ANEES: 37.8276
>>> New best VALIDATION ANEES! Saving model. <<<
-----

[DIAGNOSTIKA iter 490]
-> Rozdíl (Pred - Target): 0.967566
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi:  1.58e-03

--- Validation at iteration 495 ---
Average MSE: 8.7497, Average ANEES: 55.3410
-----

--- Iteration [500/800] ---
- Total Loss: 0.5088
- NLL: 0.3697
- Reg: 0.1391
```

```
- p1=0.345, p2=0.360

--- Validation at iteration 500 ---
Average MSE: 8.3515, Average ANEES: 41.4400
-----

[DIAGNOSTIKA iter 500]
-> Rozdíl (Pred - Target): 0.973981
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 1.54e-03

--- Validation at iteration 505 ---
Average MSE: 9.3334, Average ANEES: 43.1844
-----

--- Iteration [510/800] ---
- Total Loss: 0.6025
- NLL: 0.4646
- Reg: 0.1379
- p1=0.344, p2=0.359

--- Validation at iteration 510 ---
Average MSE: 7.4476, Average ANEES: 43.4099
-----

[DIAGNOSTIKA iter 510]
-> Rozdíl (Pred - Target): 0.973107
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 1.95e-03

--- Validation at iteration 515 ---
Average MSE: 11.0712, Average ANEES: 44.5363
-----

--- Iteration [520/800] ---
- Total Loss: 0.3815
- NLL: 0.2448
- Reg: 0.1367
- p1=0.343, p2=0.358

--- Validation at iteration 520 ---
Average MSE: 7.7661, Average ANEES: 42.0787
-----

[DIAGNOSTIKA iter 520]
-> Rozdíl (Pred - Target): 0.834144
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 3.65e-03

--- Validation at iteration 525 ---
```

```
Average MSE: 7.3099, Average ANEES: 41.6454
-----
--- Iteration [530/800] ---
- Total Loss: 0.4219
- NLL: 0.2864
- Reg: 0.1355
- p1=0.343, p2=0.357

--- Validation at iteration 530 ---
Average MSE: 12.6842, Average ANEES: 51.3034
-----

[DIAGNOSTIKA iter 530]
-> Rozdíl (Pred - Target): 0.976890
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.88e-03

--- Validation at iteration 535 ---
Average MSE: 8.6633, Average ANEES: 51.9835
-----

--- Iteration [540/800] ---
- Total Loss: 0.5178
- NLL: 0.3833
- Reg: 0.1345
- p1=0.342, p2=0.356

--- Validation at iteration 540 ---
Average MSE: 9.0310, Average ANEES: 41.8941
-----

[DIAGNOSTIKA iter 540]
-> Rozdíl (Pred - Target): 0.908892
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.17e-03

--- Validation at iteration 545 ---
Average MSE: 8.9766, Average ANEES: 43.9004
-----

--- Iteration [550/800] ---
- Total Loss: 0.3974
- NLL: 0.2641
- Reg: 0.1333
- p1=0.341, p2=0.355

--- Validation at iteration 550 ---
Average MSE: 13.6225, Average ANEES: 43.8941
-----
```

```
[DIAGNOSTIKA iter 550]
-> Rozdíl (Pred - Target): 0.982014
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 1.98e-03

--- Validation at iteration 555 ---
Average MSE: 11.3858, Average ANEES: 40.0293
-----
--- Iteration [560/800] ---
- Total Loss: 0.4958
- NLL: 0.3636
- Reg: 0.1321
- p1=0.340, p2=0.354

--- Validation at iteration 560 ---
Average MSE: 11.9617, Average ANEES: 44.5927
-----

[DIAGNOSTIKA iter 560]
-> Rozdíl (Pred - Target): 1.014207
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.60e-03

--- Validation at iteration 565 ---
Average MSE: 17.9994, Average ANEES: 48.0156
-----
--- Iteration [570/800] ---
- Total Loss: 0.5269
- NLL: 0.3958
- Reg: 0.1311
- p1=0.339, p2=0.352

--- Validation at iteration 570 ---
Average MSE: 14.7626, Average ANEES: 41.5625
-----

[DIAGNOSTIKA iter 570]
-> Rozdíl (Pred - Target): 0.927627
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 1.77e-03

--- Validation at iteration 575 ---
Average MSE: 18.4675, Average ANEES: 46.5950
-----
--- Iteration [580/800] ---
- Total Loss: 0.4327
- NLL: 0.3027
- Reg: 0.1301
```

```
- p1=0.338, p2=0.351

--- Validation at iteration 580 ---
Average MSE: 11.3487, Average ANEES: 40.4027
-----

[DIAGNOSTIKA iter 580]
-> Rozdíl (Pred - Target): 0.877814
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.20e-03

--- Validation at iteration 585 ---
Average MSE: 7.9478, Average ANEES: 45.6012
-----

--- Iteration [590/800] ---
- Total Loss: 0.4316
- NLL: 0.3026
- Reg: 0.1290
- p1=0.337, p2=0.350

--- Validation at iteration 590 ---
Average MSE: 11.3101, Average ANEES: 42.0121
-----

[DIAGNOSTIKA iter 590]
-> Rozdíl (Pred - Target): 0.894985
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.01e-03

--- Validation at iteration 595 ---
Average MSE: 8.4094, Average ANEES: 37.0757
>>> New best VALIDATION ANEES! Saving model. <<<
-----

--- Iteration [600/800] ---
- Total Loss: 0.3507
- NLL: 0.2227
- Reg: 0.1280
- p1=0.336, p2=0.349

--- Validation at iteration 600 ---
Average MSE: 10.7332, Average ANEES: 47.9549
-----

[DIAGNOSTIKA iter 600]
-> Rozdíl (Pred - Target): 0.865811
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 2.04e-03
```

```
--- Validation at iteration 605 ---
Average MSE: 7.7715, Average ANEES: 41.8769
```

```
--- Iteration [610/800] ---
- Total Loss: 0.4826
- NLL: 0.3557
- Reg: 0.1269
- p1=0.335, p2=0.348
```

```
--- Validation at iteration 610 ---
Average MSE: 7.2468, Average ANEES: 40.6923
```

```
[DIAGNOSTIKA iter 610]
-> Rozdíl (Pred - Target): 0.822364
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 1.97e-03
```

```
--- Validation at iteration 615 ---
Average MSE: 8.9078, Average ANEES: 41.9769
```

```
--- Iteration [620/800] ---
- Total Loss: 0.3560
- NLL: 0.2302
- Reg: 0.1258
- p1=0.334, p2=0.348
```

```
--- Validation at iteration 620 ---
Average MSE: 7.2887, Average ANEES: 32.9270
>>> New best VALIDATION ANEES! Saving model. <<<
```

```
[DIAGNOSTIKA iter 620]
-> Rozdíl (Pred - Target): 0.792655
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 1.34e-03
```

```
--- Validation at iteration 625 ---
Average MSE: 5.3642, Average ANEES: 41.1049
```

```
--- Iteration [630/800] ---
- Total Loss: 0.3940
- NLL: 0.2692
- Reg: 0.1248
- p1=0.333, p2=0.346
```

```
--- Validation at iteration 630 ---
Average MSE: 7.1715, Average ANEES: 40.1929
```

```

-----
[DIAGNOSTIKA iter 630]
-> Rozdíl (Pred - Target): 0.879569
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 1.92e-03

--- Validation at iteration 635 ---
Average MSE: 6.2852, Average ANEES: 44.9471
-----
--- Iteration [640/800] ---
- Total Loss: 0.3573
- NLL: 0.2336
- Reg: 0.1237
- p1=0.332, p2=0.345

--- Validation at iteration 640 ---
Average MSE: 7.1677, Average ANEES: 43.6193
-----
```



```

[DIAGNOSTIKA iter 640]
-> Rozdíl (Pred - Target): 0.861807
-> JE TO PŘESNÁ NULA? Ne
-> Min Variance v batchi: 9.66e-04

--- Validation at iteration 645 ---
Average MSE: 9.9093, Average ANEES: 45.7307
-----
--- Iteration [650/800] ---
- Total Loss: 0.3959
- NLL: 0.2732
- Reg: 0.1227
- p1=0.331, p2=0.344

--- Validation at iteration 650 ---
```

```

KeyboardInterrupt                                     Traceback (most recent call last)
Cell In[7], line 40
    37 validation_period = 5 # Chceme validovat jednou za epochu (na konci)
    39 # 3. Volání funkce
--> 40 trained_knet = trainer.train_BayesianKalmanNet(
    41     model=state_knet,
    42     train_loader=train_loader,
    43     val_loader=val_loader,
    44     device=DEVICE,
    45

```

```

46     # Parametry iterací
47     total_train_iter=total_iterations,
48     learning_rate=5e-4,           # Opatrný LR pro BKN
49     clip_grad=1.0,              # Stabilizace gradientů
50     J_samples=10,               # Počet MC vzorků (paralelně)
51
52     # Logování a validace
53     validation_period=validation_period,
54     logging_period=10,          # Výpis loss každých 10 batchů
55
56     # Warmup
57     warmup_iterations=warmup_iterations,
58     weight_decay_=1e-5
59 )

```

```

File ~/skola/KalmanNet-main/utils/trainer.py:1849, in __
  ↪ train_BayesianKalmanNet(model, train_loader, val_loader, device,_
  ↪ total_train_iter, learning_rate, clip_grad, J_samples, validation_period,_
  ↪ logging_period, warmup_iterations, weight_decay_)
1847     y_t_val = y_meas_val_batch[:, t, :]
1848     u_t_val = u_input_val_batch[:, t, :]
-> 1849     x_filtered_t, _ = model.step(y_t_val, u_t_val)
1850     val_current_x_hats.append(x_filtered_t)
1851 val_ensemble_trajectories.append(torch.stack(val_current_x_hats, dim=1))

```

```

File ~/skola/KalmanNet-main/state_NN_models/NCLT/StateBayesianKalmanNet.py:170, in __
  ↪ in StateBayesianKalmanNetNCLT.step(self, y_t, u_t)
166 # 4. NETWORK FORWARD
167 # Kontrola hidden state před vstupem do GRU
168 self._check_tensor(self.h_prev, "Hidden State Prev")
--> 170 K_vec, h_new, regs =
  ↪ self.dnn(norm_state_inno, norm_innovation, norm_diff_state, norm_diff_obs, self.h_prev)
172 # Kontrola výstupu sítě
173 self._check_tensor(K_vec, "DNN Output (K_vec)")

```

```

File ~/.local/lib/python3.10/site-packages/torch/nn/modules/module.py:1532, in __
  ↪ Module._wrapped_call_impl(self, *args, **kwargs)
1530     return self._compiled_call_impl(*args, **kwargs) # type:_
  ↪ ignore[misc]
1531 else:
-> 1532     return self._call_impl(*args, **kwargs)

```

```

File ~/.local/lib/python3.10/site-packages/torch/nn/modules/module.py:1541, in __
  ↪ Module._call_impl(self, *args, **kwargs)
1536 # If we don't have any hooks, we want to skip the rest of the logic in
1537 # this function, and just call forward.
1538 if not (self._backward_hooks or self._backward_pre_hooks or self.
  ↪ _forward_hooks or self._forward_pre_hooks

```

```

1539         or _global_backward_pre_hooks or _global_backward_hooks
1540         or _global_forward_hooks or _global_forward_pre_hooks):
-> 1541     return forward_call(*args, **kwargs)
1543 try:
1544     result = None

File ~/skola/KalmanNet-main/state_NN_models/NCLT/DNN_BayesianKalmanNet.py:49, in DNN_BayesianKalmanNetNCLT.forward(self, state_inno, innovation, diff_state, diff_obs, h_prev)
    43 # Aplikace normalizace
    44 # normalized_input = self.input_norm(nn_input)
    45
    46 # Apply dropout on normalized data
    47 activated_input, reg1 = self.concrete_dropout1(nn_input, self.
-> input_layer)
---> 49 out_gru, h_new = self.gru(activated_input.unsqueeze(0), h_prev)
    50 out_gru_squeezed = out_gru.squeeze(0)
    52 out_final, reg2 = self.concrete_dropout2(out_gru_squeezed, self.
-> output_layer)

File ~/.local/lib/python3.10/site-packages/torch/nn/modules/module.py:1532, in Module._wrapped_call_impl(self, *args, **kwargs)
    1530     return self._compiled_call_impl(*args, **kwargs) # type: ignore[misc]
    1531 else:
-> 1532     return self._call_impl(*args, **kwargs)

File ~/.local/lib/python3.10/site-packages/torch/nn/modules/module.py:1541, in Module._call_impl(self, *args, **kwargs)
    1536 # If we don't have any hooks, we want to skip the rest of the logic in
    1537 # this function, and just call forward.
    1538 if not (self._backward_hooks or self._backward_pre_hooks or self.
-> _forward_hooks or self._forward_pre_hooks
    1539         or _global_backward_pre_hooks or _global_backward_hooks
    1540         or _global_forward_hooks or _global_forward_pre_hooks):
-> 1541     return forward_call(*args, **kwargs)
1543 try:
1544     result = None

File ~/.local/lib/python3.10/site-packages/torch/nn/modules/rnn.py:1133, in GRU.forward(self, input, hx)
    1131 self.check_forward_args(input, hx, batch_sizes)
    1132 if batch_sizes is None:
-> 1133     result =_
    <_ VF.gru(input, hx, self._flat_weights, self.bias, self.num_layers,
    1134_
    <_                                         self.dropout, self.training, self.bidirectional, self.batch_first)
    1135 else:

```

```

1136     result = _VF.gru(input, batch_sizes, hx, self._flat_weights, self.
↪bias,
1137                           self.num_layers, self.dropout, self.training, self
↪bidirectional)

```

KeyboardInterrupt:

```

[ ]: # import gc
# import torch

# # === GPU/CUDA Cleanup Helper (Models Only) ===
# # Frees GPU memory by moving neural nets to CPU, deleting their references,
# # forcing GC, and clearing CUDA cache. Does NOT remove DataLoaders or trainer.

# def _cuda_mem_info():
#     if torch.cuda.is_available():
#         alloc = torch.cuda.memory_allocated() / 1e6
#         reserv = torch.cuda.memory_reserved() / 1e6
#         print(f"CUDA mem -> Allocated: {alloc:.1f} MB | Reserved: {reserv:.
↪1f} MB")
#     else:
#         print("CUDA not available.")

# print("Before cleanup:")
# _cuda_mem_info()

# # Release neural network models only
# for name in ["state_knet", "trained_model_bkn"]:
#     if name in globals():
#         obj = globals()[name]
#         try:
#             obj.to("cpu")
#         except Exception:
#             pass
#         # Drop common optimizer/scheduler refs if attached to the model
#         for attr in ["optimizer", "opt", "optim", "scheduler"]:
#             if hasattr(obj, attr):
#                 try:
#                     delattr(obj, attr)
#                 except Exception:
#                     pass
#         del globals()[name]
#         print(f"Released model '{name}' and deleted reference")

# # Force Python GC and clear CUDA cache
# gc.collect()
# if torch.cuda.is_available():

```

```

#     try:
#         torch.cuda.synchronize()
#     except Exception:
#         pass
#     torch.cuda.empty_cache()
#     try:
#         torch.cuda.ipc_collect()
#     except Exception:
#         pass

# print("After cleanup:")
# _cuda_mem_info()

```

```

[9]: if True:
    import torch
    import os

    # Protože jsi trénink přerušil, nemáme slovník s metrikami automaticky.
    # Musíme uložit přímo model 'state_knet'.

    # 1. Definuj cestu (Metriky si do názvu musíš dopsat ručně podle toho, co
    #   jsi viděl v logu naposledy,
    #   nebo použij obecný název, abys o model nepřišel).
    save_path = 'best_EKN_test_results.pth'
    # Pokud si pamatuješ hodnoty z logu (např. ANEES 23.5), můžeš je tam dopsat
    #   ručně:
    #   save_path = 'best_kalmannet_nclt_sensor_fusion_ANEES23.57_MSE8.
    #   17_interrupted.pth'

    # 2. Uložení
    # DŮLEŽITÉ: Voláme .state_dict() přímo na objektu state_knet (nebo
    #   trained_knet, pokud jsi ho přiřadil)
    # Nepoužíváme závorky ['final_model'], protože state_knet UŽ JE ten model.
    torch.save(state_knet.state_dict(), save_path)

    print(f"\n Záchrana modelu úspěšná! Uloženo do: {save_path}")

if False:
    import os
    trained_knet = state_knet
    # Přístup k metrikám ve slovníku je správně (přes závorky [])
    print(f"    -> Best ANEES: {trained_knet['best_val_anees']:.4f}")
    print(f"    -> Best MSE:    {trained_knet['best_val_mse']:.4f}")
    save_path =
        f'best_kalmannet_nclt_sensor_fusion_ANEES{trained_knet["best_val_anees"]:.4f}_MSE{trained_knet["best_val_mse"]:.4f}_nejlepsi_test_vysledky.pth '

```

```

# OPRAVA: Musíš vytáhnout model ze slovníku pomocí klíče 'final_model'
# A teprve na něm zavolat .state_dict()
torch.save(trained_knet['final_model'].state_dict(), save_path)

print(f"\n Model úspěšně uložen do: {save_path}")

```

Záchrana modelu úspěšná! Uloženo do: best_BKN_test_results.pth

```

[8]: import torch
import torch.nn.functional as F
import numpy as np
import matplotlib.pyplot as plt
import Filters

# =====
# 0. KONFIGURACE A MODEL
# =====
DT_SEC = 1.0
J_SAMPLES = 50 # 50 vzorků pro kvalitní odhad distribuce

if hasattr(sys_model, 'dt'):
    sys_model.dt = DT_SEC
    print(f"INFO: Nastaveno sys_model.dt = {DT_SEC} s")

# Načtení Bayesian modelu
try:
    trained_model_bkn = state_knet
    trained_model_bkn.eval()
    print(f"INFO: Používám Bayesian KalmanNet s J={J_SAMPLES} vzorky.")
except NameError:
    raise NameError("Chyba: 'state_knet' (BKN) neexistuje.")

from Filters import NCLT
# Inicializace klasických filtroù
ukf_filter = NCLT.UnscentedKalmanFilterNCLT(sys_model)
ekf_filter = NCLT.ExtendedKalmanFilterNCLT(sys_model)

# --- ROBUSTNÍ FUNKCE PRO ANEES ---
def calculate_anees(x_true, x_est, P_est):
    """
    Vypočítá ANEES. Očekává vstupy už se shodnou dimenzí.
    Používá pinv pro stabilitu.
    """
    T = x_true.shape[0]
    anees_list = []
    error = x_true - x_est

```

```

for t in range(T):
    e_t = error[t].unsqueeze(1) # [n, 1]
    P_t = P_est[t] # [n, n]

    try:
        P_inv = torch.linalg.pinv(P_t, hermitian=True)
    except RuntimeError:
        P_inv = torch.eye(P_t.shape[0], device=P_t.device)

    nees = torch.mm(torch.mm(e_t.t(), P_inv), e_t).item()
    anees_list.append(nees)

return np.mean(anees_list)

# =====
# 1. EVALUACE A VIZUALIZACE
# =====

results = {
    'GPS_Sensor': [], 'GPS_Authors': [],
    'EKF_MSE': [], 'EKF_ANEES': [],
    'UKF_MSE': [], 'UKF_ANEES': [],
    'BKN_MSE': [], 'BKN_ANEES': []
}

print(f"\nSpouštím detailní evaluaci na {len(test_data_raw)} trajektoriích...")

# HLAVIDKA DETAILNÍ TABULKY
print("\n" + "-" * 95)
print(f"{'TRAJEKTORIE':<12} | {'BKN MSE':<10} | {'BKN ANEES':<10} | {'EKF MSE':<10} | {'EKF ANEES':<10} | {'GPS MSE':<10}")
print("-" * 95)

for i, traj in enumerate(test_data_raw):
    # --- PRÍPRAVA DAT ---
    gt_raw = traj['ground_truth'].float().to(DEVICE)
    gps_filtered = traj['filtered_gps'].float().to(DEVICE)
    gps_filled = traj['gps'].float().to(DEVICE)
    imu_raw = traj['imu'].float().to(DEVICE)
    odo_raw = traj['filtered_wheel'].float().to(DEVICE)

    T_len = gt_raw.shape[0]

    # Input vector
    u_full = torch.stack((
        torch.nan_to_num(odo_raw[:, 0], nan=0.0),
        torch.nan_to_num(odo_raw[:, 1], nan=0.0),

```

```

    imu_raw[:, 2],
    imu_raw[:, 3]
), dim=1).to(DEVICE)

# Init State (bereme z GT)
# x_true má dimenzi 3 (X, Y, Theta)
x_true = gt_raw[:, :3]

# Init pro filtry (plná dimenze modelu)
m = sys_model.state_dim
x0_vec = torch.zeros(m).to(DEVICE)
x0_vec[0] = x_true[0, 0]; x0_vec[1] = x_true[0, 1]
if m >= 3 and x_true.shape[1] >= 3: x0_vec[4] = x_true[0, 2]

# --- A. BĚH BAYESIAN KALMANNET ---
trained_model_bkn.train() # Nutné pro MC Dropout

batch_x0 = x0_vec.unsqueeze(0).repeat(J_SAMPLES, 1)
trained_model_bkn.reset(batch_size=J_SAMPLES, initial_state=batch_x0)

bkn_samples = []
bkn_samples.append(batch_x0.unsqueeze(1))

with torch.no_grad():
    for t in range(1, T_len):
        y_t = gps_filtered[t].unsqueeze(0).repeat(J_SAMPLES, 1)
        u_t = u_full[t].unsqueeze(0).repeat(J_SAMPLES, 1)
        x_est_j, _ = trained_model_bkn.step(y_t, u_t)
        bkn_samples.append(x_est_j.unsqueeze(1))

trained_model_bkn.eval()
bkn_ensemble = torch.cat(bkn_samples, dim=1) # [J, T, m]

# Statistiky BKN
x_bkn_mean = torch.mean(bkn_ensemble, dim=0)

# Výpočet kovariance BKN
centered = bkn_ensemble - x_bkn_mean.unsqueeze(0)
centered_perm = centered.permute(1, 2, 0)
P_bkn = torch.bmm(centered_perm, centered_perm.transpose(1, 2)) / ↵
(J_SAMPLES - 1)
P_bkn = P_bkn + torch.eye(m, device=DEVICE).unsqueeze(0) * 1e-6

# --- B. EKF & UKF ---
def run_filter(flt):
    try:

```

```

        res = flt.process_sequence(gps_filtered, u_seq=u_full, Ex0=x0_vec, u
↪P0=sys_model.P0)
        return res['x_filtered'], res.get('P_filtered', torch.eye(m, u
↪device=DEVICE).repeat(T_len, 1, 1))
    except:
        return torch.zeros(T_len, m).to(DEVICE), torch.eye(m, u
↪device=DEVICE).repeat(T_len, 1, 1)

x_ekf, P_ekf = run_filter(ekf_filter)
x_ukf, P_ukf = run_filter(ukf_filter)

# --- VYHODNOCENÍ (SLICING) ---
eval_dim = x_true.shape[1] # 3 (X, Y, Theta)

def evaluate_metrics(est_x, est_P):
    # MSE (poloha X, Y)
    mse = F.mse_loss(est_x[:, :2], x_true[:, :2]).item()
    # ANEES (Oříznuté na eval_dim)
    est_x_sliced = est_x[:, :eval_dim]
    est_P_sliced = est_P[:, :eval_dim, :eval_dim]
    anees = calculate_anees(x_true, est_x_sliced, est_P_sliced)
    return mse, anees

mse_ekf, anees_ekf = evaluate_metrics(x_ekf, P_ekf)
mse_ukf, anees_ukf = evaluate_metrics(x_ukf, P_ukf)
mse_bkn, anees_bkn = evaluate_metrics(x_bkn_mean, P_bkn)

valid_mask = ~torch.isnan(gps_filtered[:, 0])
mse_gps = F.mse_loss(gps_filtered[valid_mask], x_true[valid_mask, :2]).
↪item() if valid_mask.sum() > 0 else np.nan

# Ukládání
results['GPS_Sensor'].append(mse_gps)
results['EKF_MSE'].append(mse_ekf); results['EKF_ANEES'].append(anees_ekf)
results['UKF_MSE'].append(mse_ukf); results['UKF_ANEES'].append(anees_ukf)
results['BKN_MSE'].append(mse_bkn); results['BKN_ANEES'].append(anees_bkn)

# --- VÝPIS ŘÁDKU TABULKY ---
print(f"{i+1:<12} | {mse_bkn:<10.2f} | {anees_bkn:<10.2f} | {mse_ekf:<10.
↪2f} | {anees_ekf:<10.2f} | {mse_gps:<10.2f}")

# --- VIZUALIZACE ---
gt_np = x_true.cpu().numpy()
bkn_np = x_bkn_mean.cpu().numpy()
std_x_bkn = torch.sqrt(P_bkn[:, 0, 0]).cpu().numpy()
t_axis = np.arange(T_len)

```

```

fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(16, 6))

# Graf 1: 2D Trajektorie
ax1.plot(gt_np[:, 0], gt_np[:, 1], 'k-', lw=2, label='Ground Truth')
ax1.plot(x_ekf.cpu().numpy()[:, 0], x_ekf.cpu().numpy()[:, 1], 'r--', lw=1.5, label='EKF')
ax1.plot(bkn_np[:, 0], bkn_np[:, 1], 'b-.', lw=1.5, label='BKN')
ax1.set_title(f'Traj {i+1}: 2D Position')
ax1.set_xlabel('East [m]'); ax1.set_ylabel('North [m]')
ax1.legend(); ax1.grid(True); ax1.axis('equal')

# Graf 2: Time Series X s Confidence Tube
ax2.plot(t_axis, gt_np[:, 0], 'k-', lw=1, label='GT X')
ax2.plot(t_axis, bkn_np[:, 0], 'b-', lw=1, label='BKN X')
ax2.fill_between(t_axis,
                  bkn_np[:, 0] - 3*std_x_bkn,
                  bkn_np[:, 0] + 3*std_x_bkn,
                  color='blue', alpha=0.2, label='3$\sigma$ Uncertainty')
ax2.set_title(f'X-Coord Uncertainty (ANEES: {anees_bkn:.2f})')
ax2.set_xlabel('Time Step'); ax2.set_ylabel('X Position [m]')
ax2.legend(); ax2.grid(True)

plt.tight_layout()
plt.show()

# =====
# 2. FINÁLNÍ SOUHRN
# =====

print("\n" + "="*80)
print(f"PRŮMĚRNÉ VÝSLEDKY ({len(test_data_raw)} trajektorií)")
print("=". * 80)
print(f"{'Metoda':<10} | {'MSE':<12} | {'RMSE':<12} | {'ANEES':<12}")
print("-" * 80)
print(f"{'BKN':<10} | {np.nanmean(results['BKN_MSE']):<12.2f} | {np.sqrt(np.
    nanmean(results['BKN_MSE'])):<12.2f} | {np.nanmean(results['BKN_ANEES']):<12.
    2f}")
print(f"{'EKF':<10} | {np.nanmean(results['EKF_MSE']):<12.2f} | {np.sqrt(np.
    nanmean(results['EKF_MSE'])):<12.2f} | {np.nanmean(results['EKF_ANEES']):<12.
    2f}")
print("=". * 80)

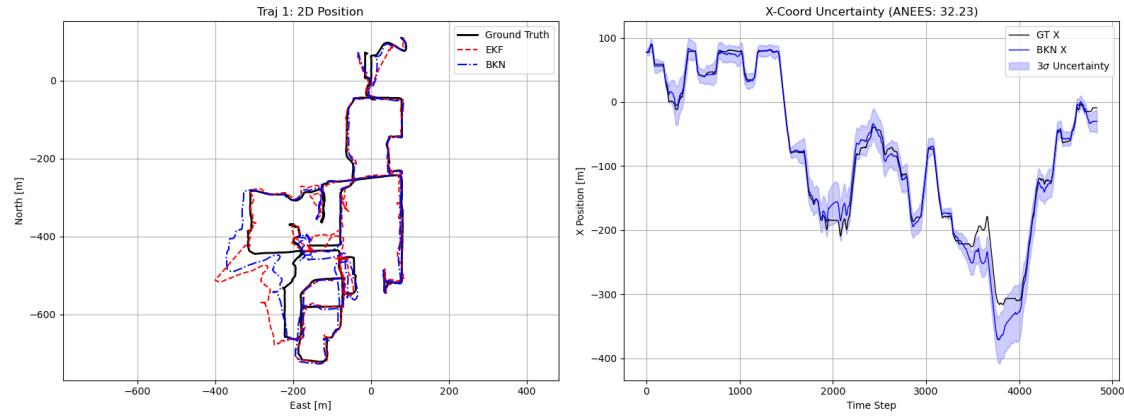
```

INFO: Nastaveno sys_model.dt = 1.0 s
 INFO: Používám Bayesian KalmanNet s J=50 vzorky.

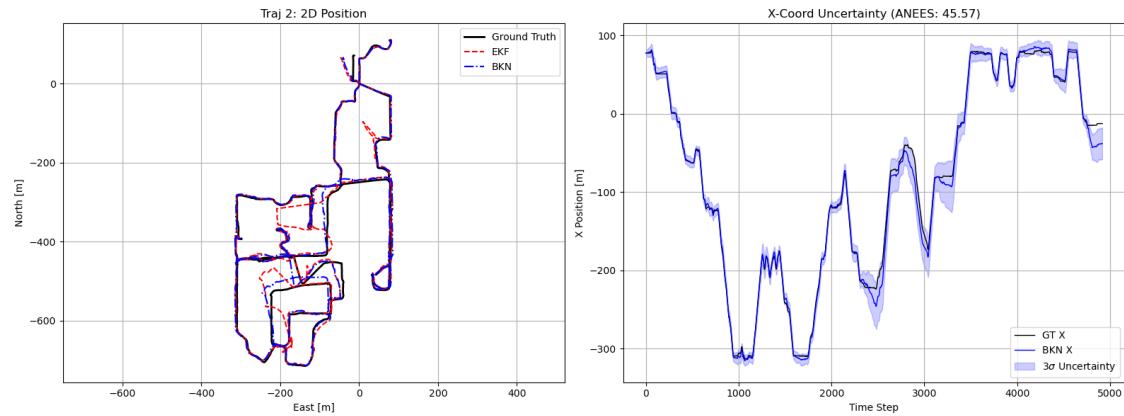
Spouštím detailní evaluaci na 3 trajektoriích..

TRAJEKTORIE | BKN MSE | BKN ANEES | EKF MSE | EKF ANEES | GPS MSE

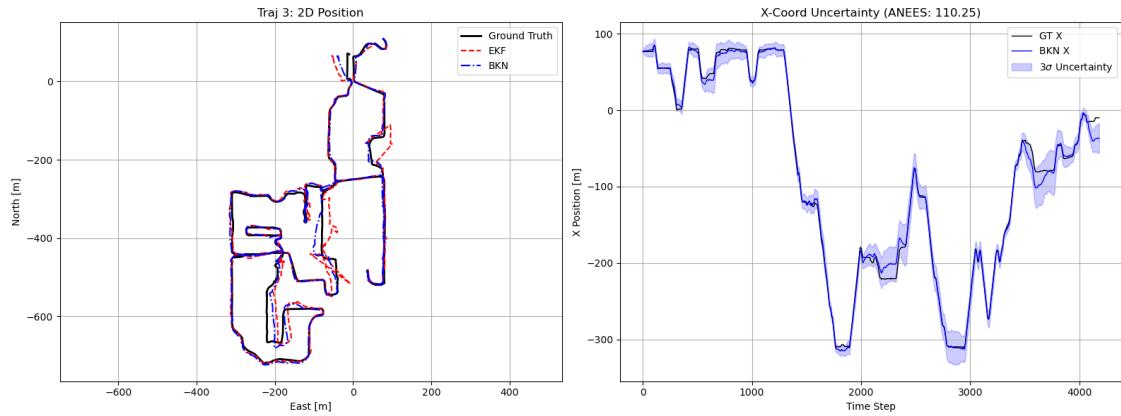
1 | 185.61 | 32.23 | 443.11 | 1027.02 | 164.46



2 | 44.76 | 45.57 | 221.68 | 445.10 | 51.13



3 | 34.27 | 110.25 | 128.44 | 311.25 | 39.94



PRŮMĚRNÉ VÝSLEDKY (3 trajektorií)

Metoda	MSE	RMSE	ANEES
BKN	88.21	9.39	62.68
EKF	264.41	16.26	594.46
