

# Hands on - Digital Twin of a low field scanner

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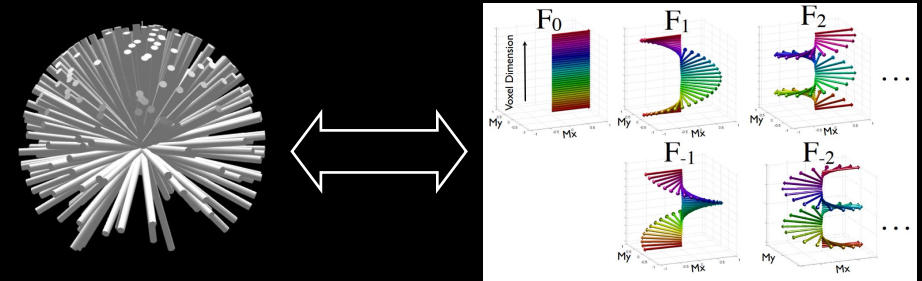


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# Overview

You know by now:

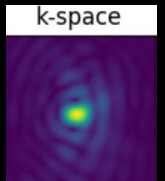
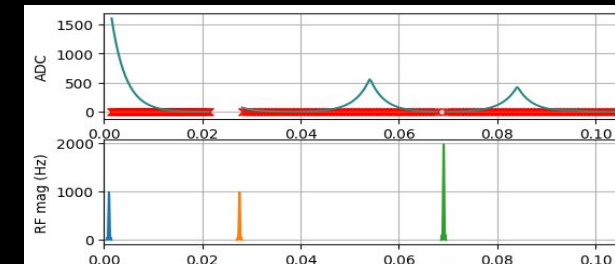
- That we can simulate MRI & most efficient with EPG/PDG simulations
- You can simulate Pulseq/PyPulseq definitions directly in Python/Colab



```
signal = mr0.execute_graph(graph, seq0, obj_p)
```

What we will do now

- Adapt the simulation
  - **For a low field system** including typical phantom parameters
  - **For a 3D sequence.** Required for SNR-reasons, but long and slow in simulation.
- Explore the low field system and 3D sequence



# Exploring is playing

“Play is the highest form of research.” – Albert Einstein

while playing at the MR scanner is the best way to learn,  
it can be slow and can be restricted by hardware/software

A playful simulation allows you

- ..to test many new ideas in short time,
- ..to ask the questions „what if“ with less limits
- ..to identify the most interesting things to „play“ out at the scanner

# General steps

An MR simulation requires four ingredients:

1. a defined MR sequence
2. a defined MR phantom with all tissue and system properties
3. a call of the simulator with sequence and phantom
4. a reconstruction of k-space signals to images

# General steps

The MR simulation requires four ingredients:

Load Pulseseq MTSE , ....

Simpler seqs in MR0 playground

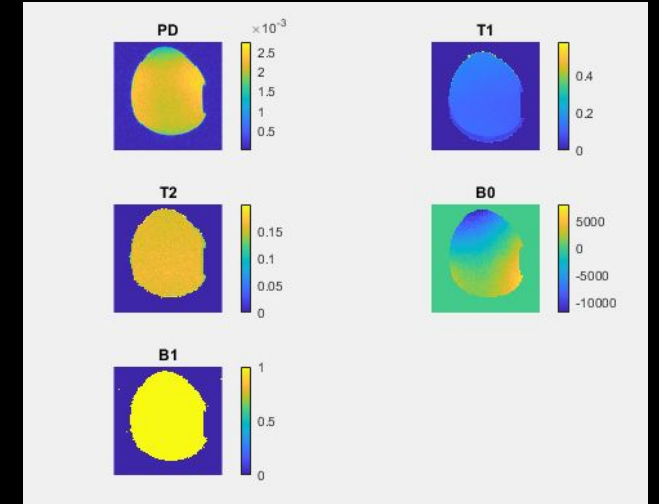
1. a defined MR sequence
2. a defined MR phantom with all tissue and system properties
3. a call of the simulator with sequence and phantom
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# ingredient #2

## 2. Phantom and system definition

- quantitative maps: PD, T1, T2, (T2', D)
- system-dependent maps: B0, B1+, (B1-)
- store in matlab or python, load:

```
○ import MRzeroCore as mr0
○ obj_p =
  mr0.VoxelGridPhantom.load_mat('low_field_phantom.mat')
```



In the colab script, go to section S3.



Inhalt



MaRCoS system optimized 3D MTSE + MR-zero simulation



Simulation:



Exercises or Games - Explore Simulation:



Code Sections for 3D MTSE forMaRCoS in a low field phantom

S0. environment setup (just need to run it one time)

S1. Pulseseq sequence, wrapped as a function  
create\_mtse

Run all below wrapper

S2. interactive changes on seq

**S3. Load phantom and manipulate**

S4. Simulate

S5. Explore Interactive seq.plot() - experimental

S6. Reconstruction and image display

Matlab Code for generating phantoms in MR-zero

+ Bereich



+ Code + Text



T4

RAM  
Laufwerk

Gemini



## S3. Load phantom and manipulate



```
#@title S3. Load phantom and manipulate
dB0 = 0
phantom_fn = "low_field_phantom.mat" # @param ["low_field_phantom.mat"]
slice_select = [8] # @param {type:"raw"}
sz = [nRD, nPH, 1]
# %% S4: SETUP SPIN SYSTEM/object on which we can
obj_p = mr0.VoxelGridPhantom.load_mat(phantom_fn)
obj_p.T2dash[:] = 30e-3
obj_p.D *= 0

#obj_p = obj_p.interpolate(40, 40, 16) # all slices
#obj_p = obj_p.interpolate(40, 40, 16).slices(slice_select)
obj_p = obj_p.interpolate(40, 40, 16).slices(slice_select)

# change the size in SI untis (m), here 15 cm
obj_p.size = torch.tensor([0.15,0.15,0.15])

# Manipulate loaded data
obj_p.B0*= 1/40 # alter the B0 inhomogeneity
rB1_factor= 1 # @param {type: "slider", min: 0, max: 1}
dB0_factor= 1 # @param {type: "slider", min: 0, max: 1}
T1_factor= 1 # @param {type: "slider", min: 0, max: 1}
T2_factor= 1 # @param {type: "slider", min: 0, max: 1}
T2dash_factor= 1 # @param {type: "slider", min: 0, max: 1}
D = 0 # @param {type: "slider", min: 0, max: 5}
obj_p.T1*= T1_factor
obj_p.T2*= T2_factor
obj_p.T2dash*= T2dash_factor
obj_p.B1*= rB1_factor
obj_p.B0*= dB0_factor
obj_p.D[:]= D
obj_p.plot()
```

phantom\_fn: low\_field\_phantom.mat

slice\_select: [8]

rB1\_factor: 1

dB0\_factor: 1

T1\_factor: 1

T2\_factor: 1

T2dash\_factor: 1

D: 0

An single slice phantom, does not change the simulation outcome (except SNR), as the 3D sequence dynamic is unchanged.

-> valid speedup

# To be able to play fast, we need speed-ups

Speed up “phantom” - less voxels to simulate

## Speedup sequence

80x80x8 x4 -> 40x40x2 x4

## Speed up simulation

-> accuracy

-> GPU



# To be able to play fast, we need speed-ups

## Speed up phantom

## Speedup sequence

80x80x8 x4 -> 40x40x2 x4

## Speed up simulation

-> accuracy

-> GPU

▼ S2. interactive changes on seq

seq=create\_mtse(

FAex:  90

FAref:  180

fsp\_r:  1

nRD:  40

nPH:  40

n3D:  2

n\_echo:  4

TE:

TR:

rf\_ex\_phase:  90

t0:

t1:

)

# To be able to play fast, we need speed-ups

## Speed up phantom

## Speedup sequence

80x80x8 x4- $\rightarrow$  40x40x2 x4

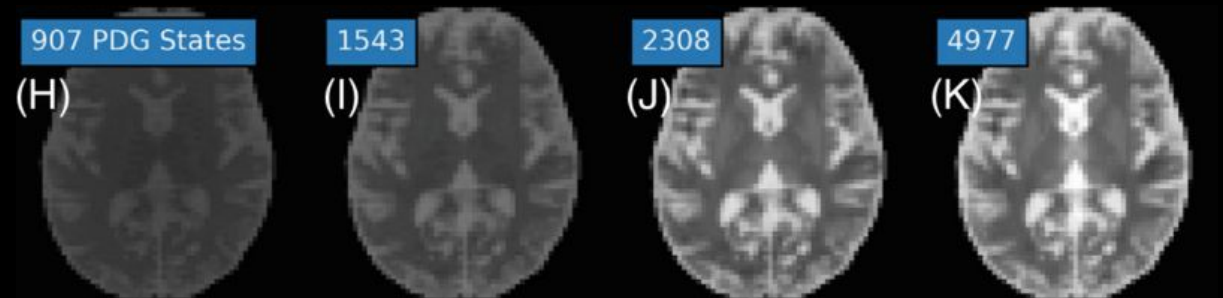
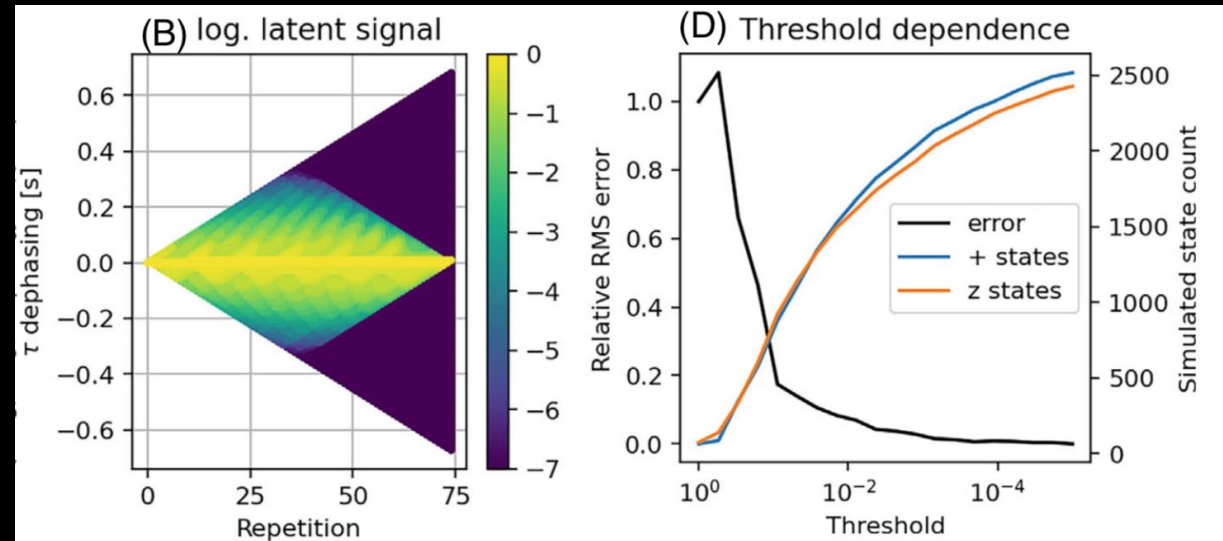
## Speed up simulation

-> accuracy (Malik et al. Equivalence of EPG and Isochromats, ISMRM 2016)

```
signal = mr0.execute_graph(graph, seq0, obj_p)
```

```
#Faster simulation with lower accuracy:
```

```
signal = mr0.execute_graph(graph, seq0, obj_p, min_emitted_signal=0.05, min_latent_signal=0.05)
```



# To be able to play fast, we need speed-ups

## Speed up phantom

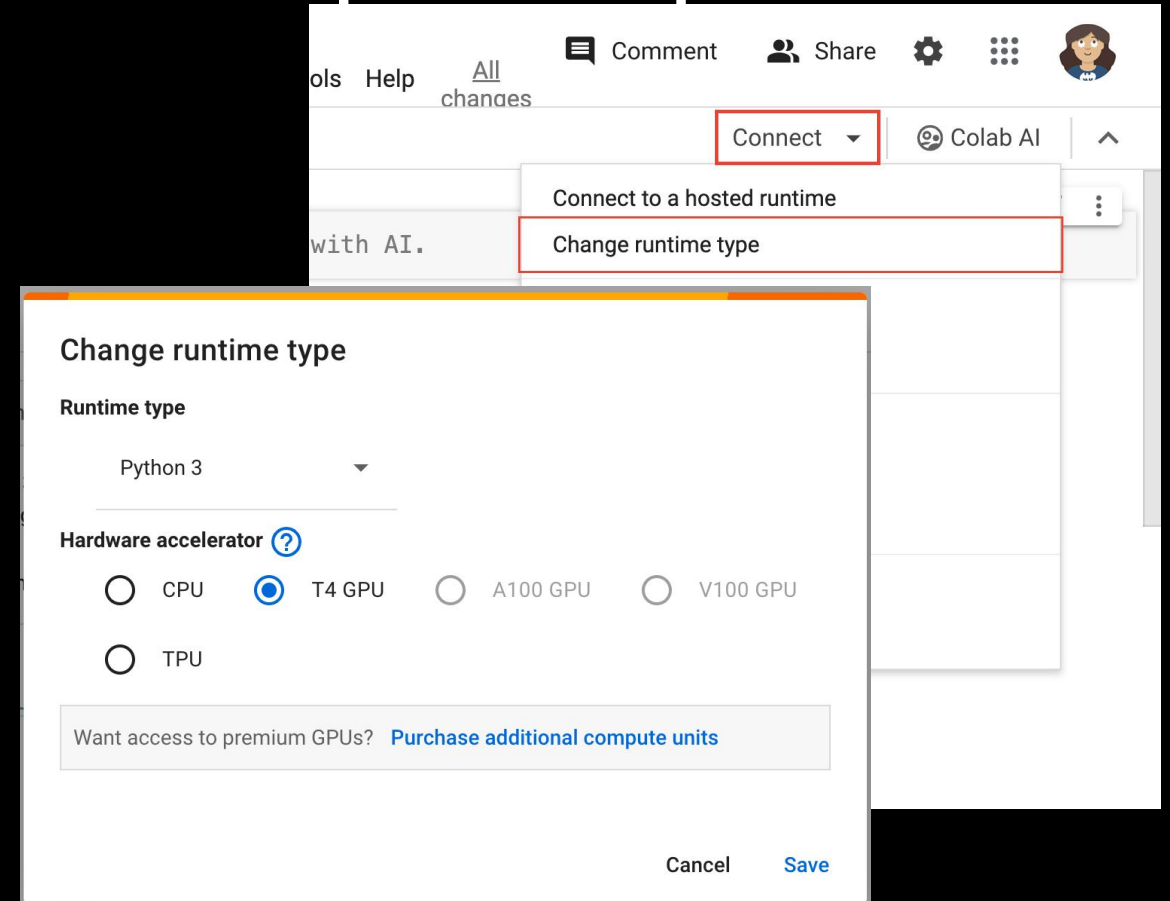
## Speedup sequence

80x80x8 x4-> 40x40x2 x4

## Speed up simulation

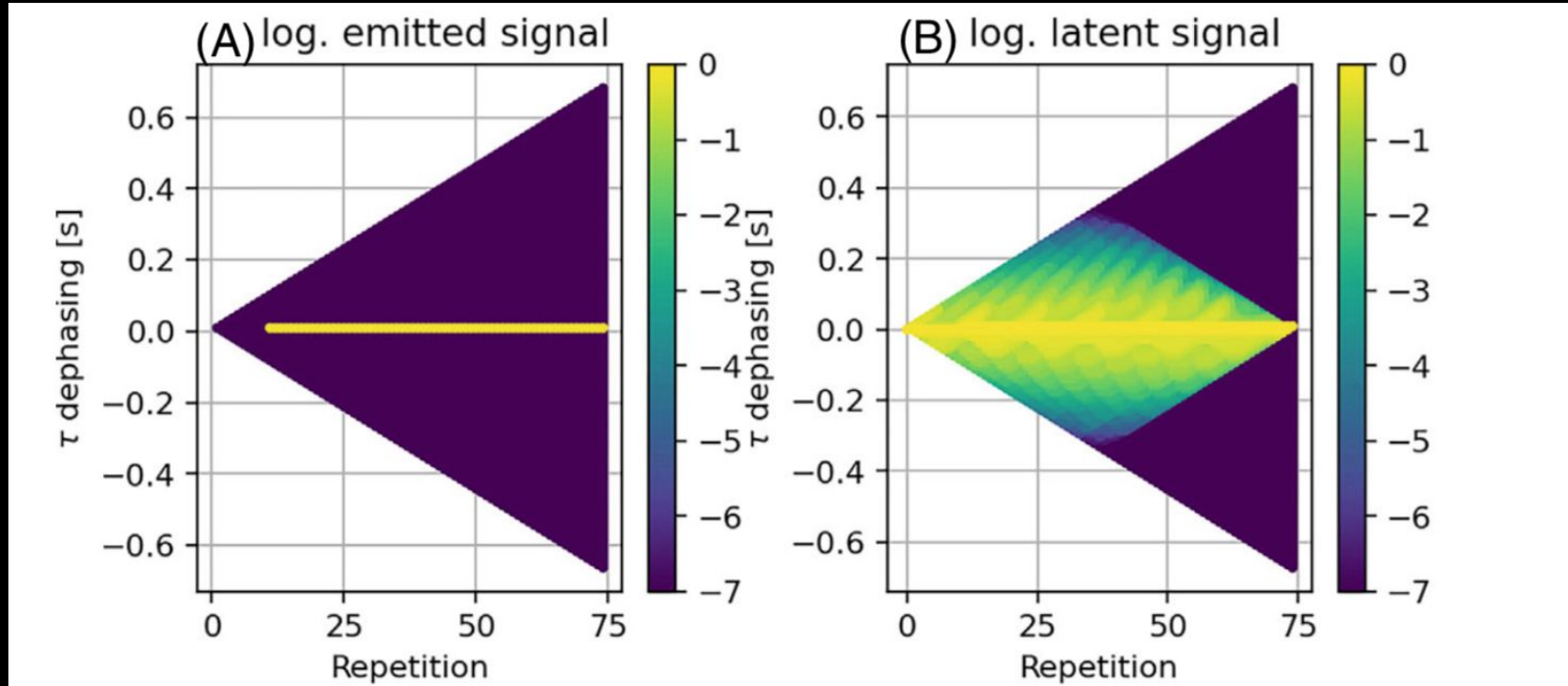
-> accuracy

-> GPU



```
signal=mr0.execute_graph(graph,  
seq0.cuda(), obj_p.cuda())
```

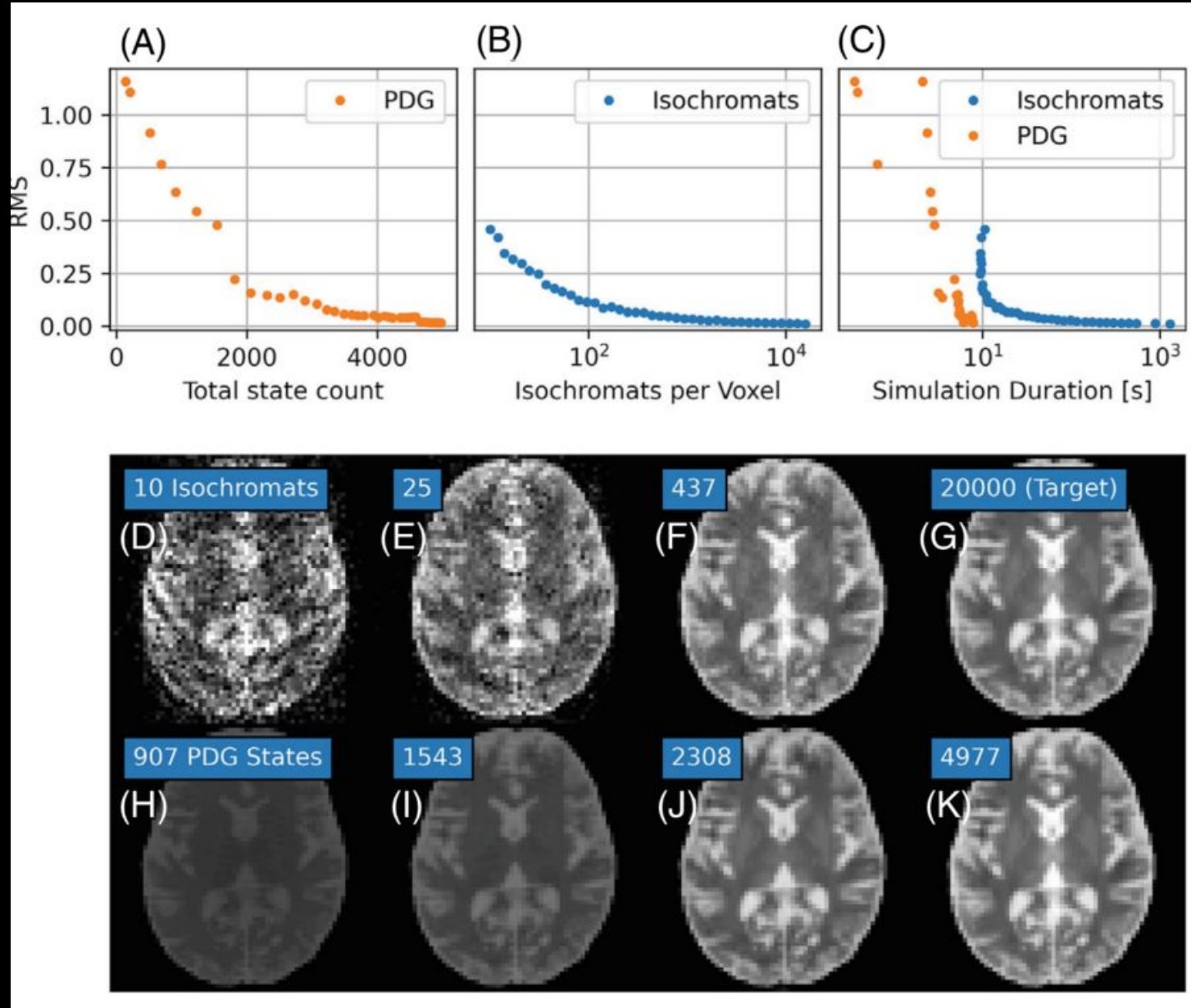
# State selection and latent signal



Emitted signal:  
Contribution of the state  
to the current signal

Latent signal:  
Contribution of the state to  
any later acquired signal

# Time efficiency

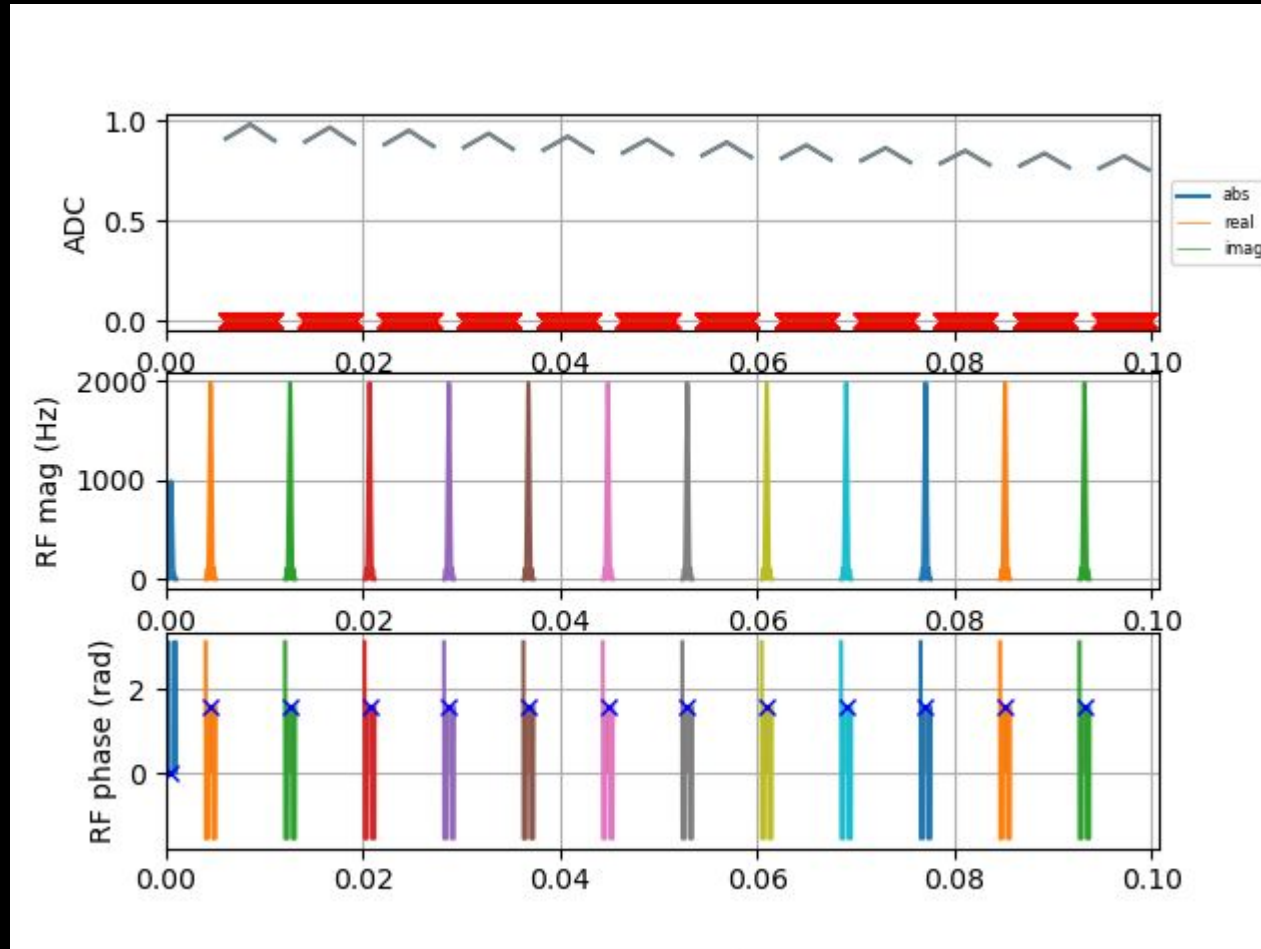


Interesting behavior:

- No Monte-Carlo noise, but missing signal
- Flexible threshold
- Faster with more diffusion, lower T2, shorter sequences
- “Regular” sequences are “detected” ( $\text{EPG} \subseteq \text{PDG}$ )
- Sequences with large, but dephased states benefit strongly from PDG

# EPG $\subseteq$ PDG and PDG $>$ EPG for a simple example

- EPG is most famous for describing TSE trains like this one:

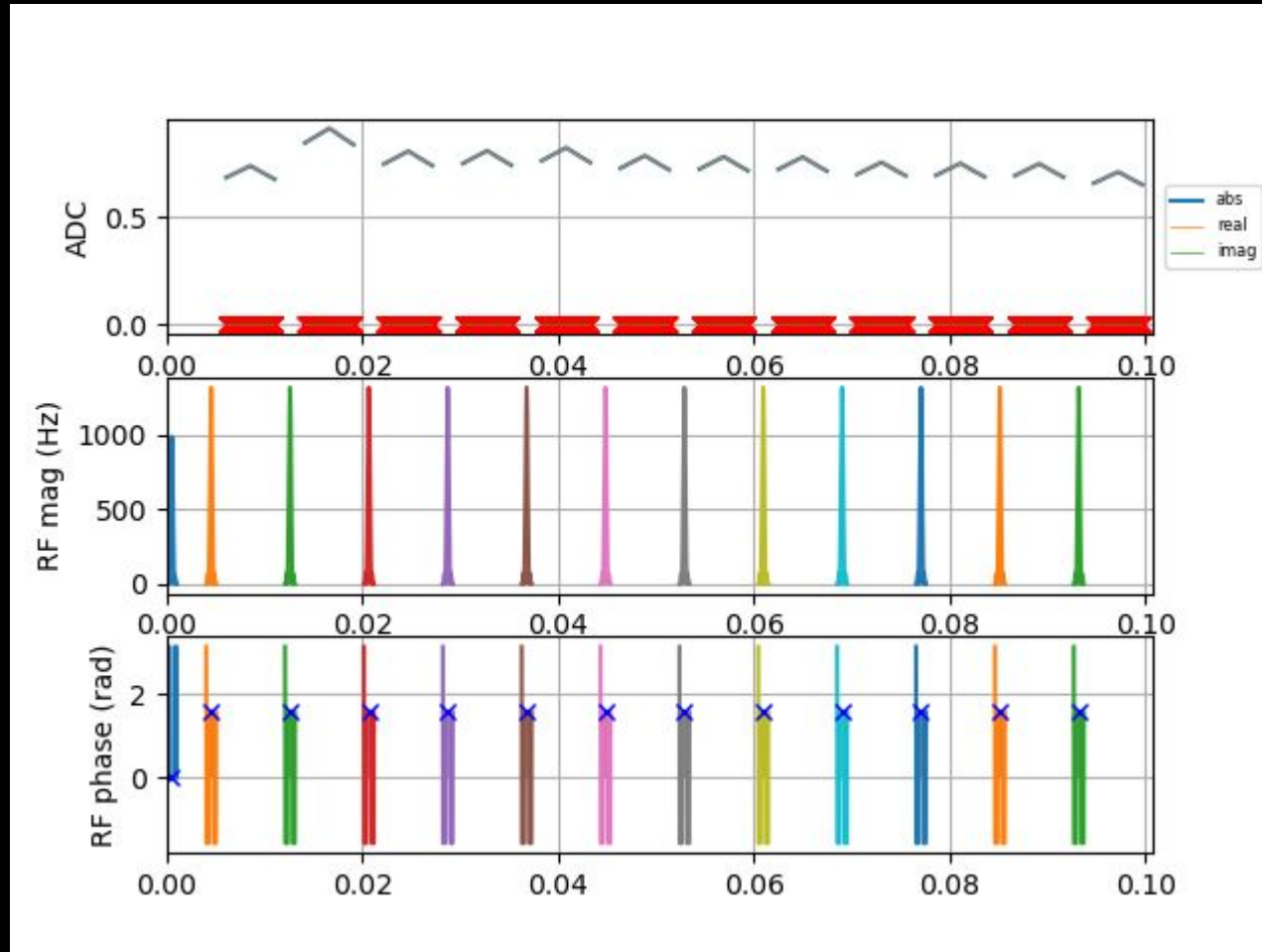


For this 90° -180° case this is an exponential decay, but EPG can also describe e.g. 90°-120° TSEs, where stimulated echoes contribute.



# EPG $\subseteq$ PDG and PDG $>$ EPG for a simple example

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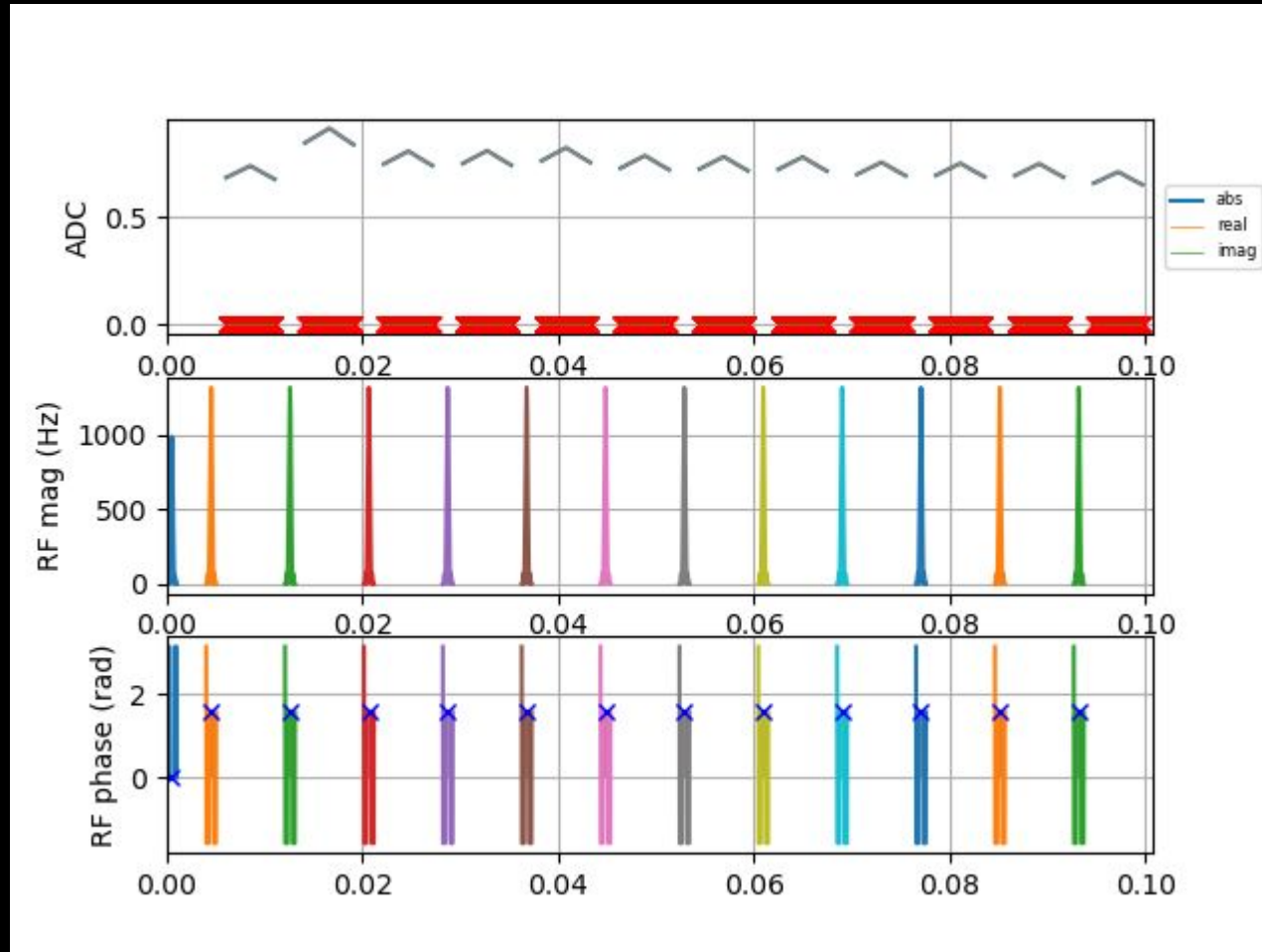


90°-120° TSE:  
Stimulated echo contribution  
changes exp decay

-> described by EPG

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90°-120° TSE:  
Stimulated echo contribution  
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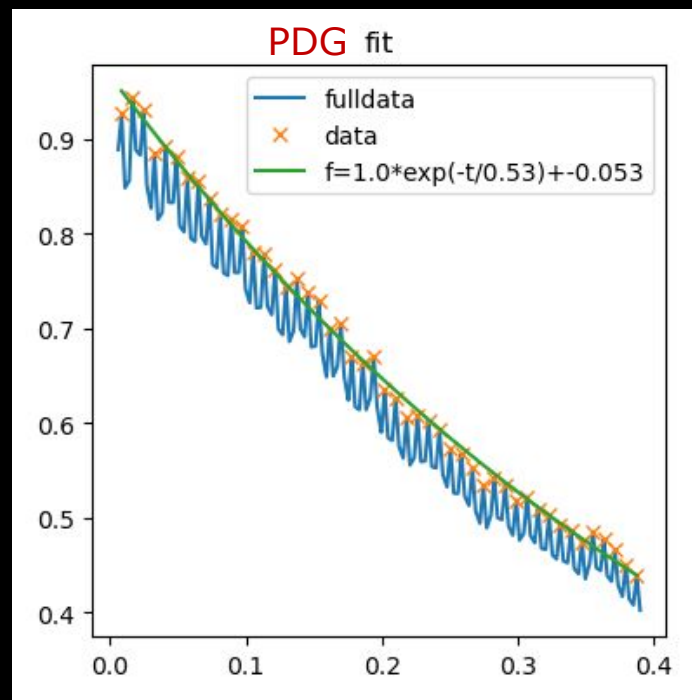
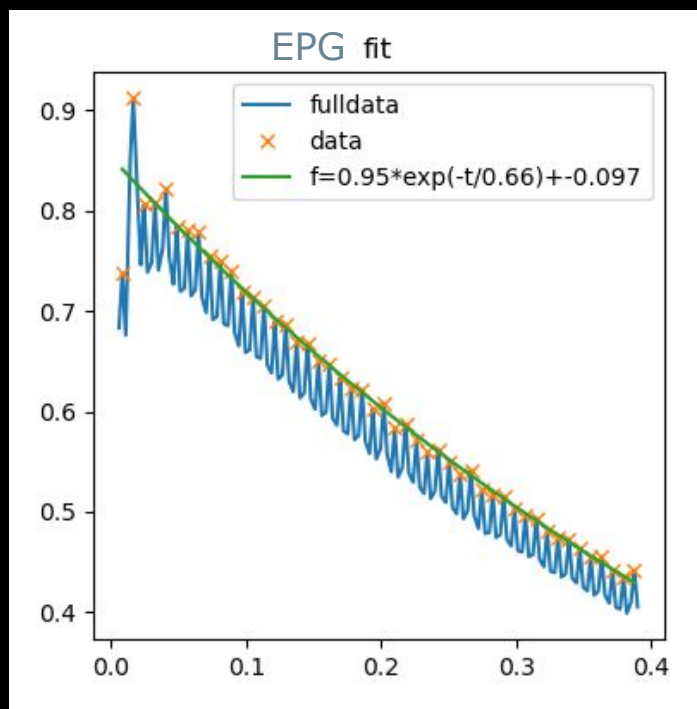
-> described by EPG

Let's look at the exact same  
sequence with PDG!



# EPG $\subseteq$ PDG and PDG $>$ EPG for a simple example

- This even leads to a different decay rate!



90°-120° TSE:  
Stimulated echo contribution  
changes exp decay

-> described by EPG

FID + SE contribution  
Also changes exp decay

-> not described by EPG

-> described by PDG