
Statespace

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CONTENTS:

1	install	3
2	processors	5
2.1	kalman.py	5
2.2	sigmapoint.py	5
2.3	particle.py	6
3	models	7
3.1	basemodel.py	7
3.2	onestate.py	7
3.3	threestate.py	7
3.4	bearingsonly.py	8
Python Module Index		9
Index		11

this project focuses on reference problems in kalman, sigma-point, and particle processors - in particular, numpy and eigen implementations, using classic vector and matrix representations so things look like what's in the literature.

**CHAPTER
ONE**

INSTALL

have switched to a container build environment - reproducible using the dockerfile. everything needed is there and can be reproduced for a local environment. docker image is available on docker hub.

CHAPTER
TWO

PROCESSORS

2.1 kalman.py

baseline extended kalman filters. both the standard textbook form, and the ud factorized form.

```
class statespace.processors.kalman(*args, **kwargs)
    classical extended kalman filter
    ekf(model)
        basic form
    ekfud(model)
        UD factorized form
    observational(xin, Uin, Din, H, obs, R, yhat)
        bierman observation update
    temporal(xin, Uin, Din, Phi, Gin, Q)
        thornton temporal update
    udfactorize(M)
        UD factorization
```

2.2 sigmapoint.py

sigma-point sampling kalman filters. this is the ukf or unscented kalman filter, or ‘modern’ kalman filtering. it’s essentially somewhere between a classical kalman filter, where uncertainty is represented as gaussian, and a particle filter, where uncertainty has arbitrary shape. here uncertainty is deterministically sampled at a small number of points, the sigma points or sigma particles. in a particle filter the number and role of the particles are increased.

```
class statespace.processors.sigmapoint.SigmaPoint(*args, **kwargs)
    modern sigma-point deterministic sampling kalman filter
    choldowndate(R, z)
        cholesky downdate
    cholupdate(R, z)
        cholesky update
    observational(x, o, X, Y, W, Xtil, Ytil, Pxy, S, Sobs)
        cholesky factorized observational update
```

```
spf(model)
    sigma-point deterministic sampling kalman filter

spfcholesky(model)
    cholesky factorized sigma-point sampling kalman filter

temporal(x, XY, W, Xtil, S, Sproc, u)
    cholesky factorized temporal update
```

2.3 particle.py

particle filters, sequential monte carlo sampling processors. sampling here is random, not deterministic as in the sigma-point processor. and the idea of resampling and growing new particles comes to the fore. the particles are random and new ones can be introduced freely.

```
class statespace.processors.particle.Particle(*args, **kwargs)
    particle filter, sequential monte carlo processor

pf(model)
    basic sampling-importance-resampling (SIR) particle filter, bootstrap particle filter, condensation particle
    filter, survival of the fittest algorithm
```

CHAPTER
THREE

MODELS

will probably be moving towards a higher-level statespace model, encompassing specific lower-level models - possibly something involving a translator / converter / adaptor... the models here are already an extremely primitive form of that - making them as similar as possible from the perspective of the classical, modern, particle processors. we can think about these becoming specific cases of something more fundamental.

3.1 basemodel.py

placeholder for what could grow to become a higher-level statespace model - with individual models inheriting and overriding.

```
class statespace.models.basemodel.BaseModel
    base model
```

3.2 onestate.py

a simple as possible one-state example with non linear temporal and observation updates. it's a common example in the candy and jazwinisky books. based on real world reentry vehicle tracking.

```
class statespace.models.onestate.Onestate
    one-state reference model
```

3.3 threestate.py

three-state extension of the the one-state model. non linear temporal and observation updates.

```
class statespace.models.threestate.Threestate
    three-state reference model
```

3.4 bearingsonly.py

the bearings only problem has some interesting history. it's basically about being on a sub. your sub is travelling along steadily and you begin hearing the sound of a ship at some bearing. over time and as the bearing changes, you can estimate the relative position and velocity of the ship. at some point you make a course change for your sub to pursue the ship.

```
class statespace.models.bearingsonly.BearingsOnly  
    bearings-only tracking problem
```

PYTHON MODULE INDEX

S

`statespace.models.basemodel`, 7
`statespace.models.bearingsonly`, 8
`statespace.models.onestate`, 7
`statespace.models.threestate`, 7
`statespace.processors.kalman`, 5
`statespace.processors.particle`, 6
`statespace.processors.sigmapoint`, 5

INDEX

B

BaseModel (*class in statespace.models.basemodel*), 7
BearingsOnly (*class in statespace.models.bearingsonly*), 8

C

choldowndate() (*statespace.processors.sigmapoint.SigmaPoint method*), 5
cholupdate() (*statespace.processors.sigmapoint.SigmaPoint method*), 5

E

ekf() (*statespace.processors.kalman.Kalman method*), 5
ekfud() (*statespace.processors.kalman.Kalman method*), 5

K

Kalman (*class in statespace.processors.kalman*), 5

M

module
 statespace.models.basemodel, 7
 statespace.models.bearingsonly, 8
 statespace.models.onestate, 7
 statespace.models.threestate, 7
 statespace.processors.kalman, 5
 statespace.processors.particle, 6
 statespace.processors.sigmapoint, 5

O

observational() (*statespace.processors.kalman.Kalman method*), 5
observational() (*statespace.processors.sigmapoint.SigmaPoint method*), 5
Onestate (*class in statespace.models.onestate*), 7

P

Particle (*class in statespace.processors.particle*), 6

pf() (*statespace.processors.particle.Particle method*), 6

S

SigmaPoint (*class in statespace.processors.sigmapoint*), 5
spf() (*statespace.processors.sigmapoint.SigmaPoint method*), 5
spfcholesky() (*statespace.processors.sigmapoint.SigmaPoint method*), 6
statespace.models.basemodel module, 7
statespace.models.bearingsonly module, 8
statespace.models.onestate module, 7
statespace.models.threestate module, 7
statespace.processors.kalman module, 5
statespace.processors.particle module, 6
statespace.processors.sigmapoint module, 5

T

temporal() (*statespace.processors.kalman.Kalman method*), 5
temporal() (*statespace.processors.sigmapoint.SigmaPoint method*), 6
Threestate (*class in statespace.models.threestate*), 7

U

udfactorize() (*statespace.processors.kalman.Kalman method*), 5