

pestpp-glm

May 1, 2019

1 PESTPP-GLM

In this notebook, we will run PESTPP-GLM in standard parameter estimation mode and regularization mode. In both cases, we will use the baked-in bayes-linear posterior monte carlo analysis to get posterior forecast PDFs. We will use the prior monte carlo outputs as the prior forecast PDF.

```
In [1]: import os
import shutil
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import flopy
import pyemu
```

flopy is installed in /Users/jeremyw/Dev/gw1876/activities_2day_mfm/notebooks/flopy

```
In [2]: t_d = "template"
m_d = "master_glm"
```

```
In [3]: pst = pyemu.Pst(os.path.join(t_d, "freyberg.pst"))
pst.write_par_summary_table(filename="none")
```

```
Out[3]:
```

	type	transform	count	initial value	\
cn_strt6	cn_strt6	log	1	0	
flow	flow	log	1	0	
cn_hk7	cn_hk7	log	1	0	
drncond_k00	drncond_k00	log	10	0	
gr_vka3	gr_vka3	log	705	0	
gr_prsity5	gr_prsity5	log	705	0	
cn_rech4	cn_rech4	log	1	0	
pp_sy0	pp_sy0	log	32	0	
pp_strt1	pp_strt1	log	32	0	
gr_sy5	gr_sy5	log	705	0	
pp_ss2	pp_ss2	log	32	0	
pp_vka0	pp_vka0	log	32	0	
cn_prsity7	cn_prsity7	log	1	0	

gr_prsity4	gr_prsity4	log	705	0
gr_vka5	gr_vka5	log	705	0
pp_hk0	pp_hk0	log	32	0
gr_sy4	gr_sy4	log	705	0
gr_ss5	gr_ss5	log	705	0
pp_hk1	pp_hk1	log	32	0
gr_rech2	gr_rech2	log	705	0
gr_strt4	gr_strt4	log	705	0
pp_prsity0	pp_prsity0	log	32	0
strk	strk	log	40	0
pp_ss1	pp_ss1	log	32	0
gr_sy3	gr_sy3	log	705	0
welflux	welflux	log	2	0 to 0.176091
pp_prsity1	pp_prsity1	log	32	0
gr_rech3	gr_rech3	log	705	0
cn_ss6	cn_ss6	log	1	0
cn_prsity6	cn_prsity6	log	1	0
...
pp_rech1	pp_rech1	log	32	0
pp_vka2	pp_vka2	log	32	0
cn_rech5	cn_rech5	log	1	-0.39794
pp_rech0	pp_rech0	log	32	0
cn_strt7	cn_strt7	log	1	0
welflux_k02	welflux_k02	log	6	0
gr_strt5	gr_strt5	log	705	0
cn_sy6	cn_sy6	log	1	0
gr_hk5	gr_hk5	log	705	0
pp_vka1	pp_vka1	log	32	0
gr_ss4	gr_ss4	log	705	0
cn_sy8	cn_sy8	log	1	0
pp_sy2	pp_sy2	log	32	0
cn_hk8	cn_hk8	log	1	0
pp_sy1	pp_sy1	log	32	0
cn_prsity8	cn_prsity8	log	1	0
cn_vka7	cn_vka7	log	1	0
cn_sy7	cn_sy7	log	1	0
cn_vka8	cn_vka8	log	1	0
cn_vka6	cn_vka6	log	1	0
pp_ss0	pp_ss0	log	32	0
pp_prsity2	pp_prsity2	log	32	0
gr_strt3	gr_strt3	log	705	0
pp_strt0	pp_strt0	log	32	0
pp_strt2	pp_strt2	log	32	0
cn_ss8	cn_ss8	log	1	0
pp_hk2	pp_hk2	log	32	0
gr_hk4	gr_hk4	log	705	0
cn_ss7	cn_ss7	log	1	0
gr_hk3	gr_hk3	log	705	0

	upper bound	lower bound	standard deviation
cn_strt6	0.0211893	-0.0222764	0.0108664
flow	0.09691	-0.124939	0.0554622
cn_hk7	1	-1	0.5
drncond_k00	1	-1	0.5
gr_vka3	1	-1	0.5
gr_prsity5	0	-1	0.25
cn_rech4	0.0791812	-0.09691	0.0440228
pp_sy0	0.243038	-0.60206	0.211275
pp_strt1	0.0211893	-0.0222764	0.0108664
gr_sy5	0.243038	-0.60206	0.211275
pp_ss2	1	-1	0.5
pp_vka0	1	-1	0.5
cn_prsity7	0	-1	0.25
gr_prsity4	0	-1	0.25
gr_vka5	1	-1	0.5
pp_hk0	1	-1	0.5
gr_sy4	0.243038	-0.60206	0.211275
gr_ss5	1	-1	0.5
pp_hk1	1	-1	0.5
gr_rech2	0.0413927	-0.0457575	0.0217875
gr_strt4	0.0211893	-0.0222764	0.0108664
pp_prsity0	0	-1	0.25
strk	2	-2	1
pp_ss1	1	-1	0.5
gr_sy3	0.243038	-0.60206	0.211275
welflux	0.176091 to 0.30103	-0.30103 to 0	0.0752575 to 0.11928
pp_prsity1	0	-1	0.25
gr_rech3	0.0413927	-0.0457575	0.0217875
cn_ss6	1	-1	0.5
cn_prsity6	0	-1	0.25
...
pp_rech1	0.0413927	-0.0457575	0.0217875
pp_vka2	1	-1	0.5
cn_rech5	-0.09691	-1	0.225772
pp_rech0	0.0413927	-0.0457575	0.0217875
cn_strt7	0.0211893	-0.0222764	0.0108664
welflux_k02	1	-1	0.5
gr_strt5	0.0211893	-0.0222764	0.0108664
cn_sy6	0.243038	-0.60206	0.211275
gr_hk5	1	-1	0.5
pp_vka1	1	-1	0.5
gr_ss4	1	-1	0.5
cn_sy8	0.243038	-0.60206	0.211275
pp_sy2	0.243038	-0.60206	0.211275
cn_hk8	1	-1	0.5
pp_sy1	0.243038	-0.60206	0.211275

cn_prsity8	0	-1	0.25
cn_vka7	1	-1	0.5
cn_sy7	0.243038	-0.60206	0.211275
cn_vka8	1	-1	0.5
cn_vka6	1	-1	0.5
pp_ss0	1	-1	0.5
pp_prsity2	0	-1	0.25
gr_strt3	0.0211893	-0.0222764	0.0108664
pp_strt0	0.0211893	-0.0222764	0.0108664
pp_strt2	0.0211893	-0.0222764	0.0108664
cn_ss8	1	-1	0.5
pp_hk2	1	-1	0.5
gr_hk4	1	-1	0.5
cn_ss7	1	-1	0.5
gr_hk3	1	-1	0.5

[65 rows x 7 columns]

1.0.1 reduce the number of adjustable parameters

This is the painful part: we cant use 10K+ pars because we cant wait around for that many runs and then the linear algebra of factoring a 10k+ by 10K+ matrix is also difficult. So that means we need to fix a lot a parameters #frownyface

```
In [4]: par = pst.parameter_data
```

```
In [5]: # grid-scale pars
gr_pars = par.loc[par.pargp.apply(lambda x: "gr" in x), "parnme"]
par.loc[gr_pars, "partrans"] = "fixed"
pst.npar_adj
```

```
Out [5]: 719
```

```
In [6]: # these are the sfr conductance parameters - Ive left all 40 adjustable
# but if you uncomment this, it will tie them into 1 parameter effectively
# strk_pars = par.loc[par.pargp=="strk", "parnme"]
# p1 = strk_pars.iloc[0]
# par.loc[strk_pars.iloc[1:], "partrans"] = "tied"
# par.loc[strk_pars.iloc[1:], "partied"] = p1
pst.npar_adj
```

```
Out [6]: 719
```

```
In [7]: par.loc[par.pargp.apply(lambda x: "pp" in x), "pargp"].unique()
```

```
Out [7]: array(['pp_hk0', 'pp_hk1', 'pp_hk2', 'pp_prsity0', 'pp_prsity1',
                'pp_prsity2', 'pp_rech0', 'pp_rech1', 'pp_ss0', 'pp_ss1', 'pp_ss2',
                'pp_strt0', 'pp_strt1', 'pp_strt2', 'pp_sy0', 'pp_sy1', 'pp_sy2',
                'pp_vka0', 'pp_vka1', 'pp_vka2'], dtype=object)
```

Fix the storage pilot points - we still have layer-scale storage pars adjustable

```
In [8]: #s_pars = par.loc[par.pargp.apply(lambda x: "pp" in x and ("ss" in x or "sy" in x)), "p"]  
        #par.loc[s_pars, "partrans"] = "fixed"  
        pst.npar_adj
```

Out[8]: 719

```
In [9]: adj_par = par.loc[par.partrans=="log", :]  
        adj_par.pargp.value_counts().sort_values()
```

```
Out[9]: cn_strt6      1  
        cn_strt8      1  
        cn_prsity7     1  
        flow          1  
        cn_ss8         1  
        cn_vka6         1  
        cn_rech4        1  
        cn_ss7          1  
        cn_sy7          1  
        cn_vka8         1  
        cn_vka7         1  
        cn_prsity8      1  
        cn_rech5        1  
        cn_hk8          1  
        cn_hk7          1  
        cn_sy8          1  
        cn_sy6          1  
        cn_strt7        1  
        cn_hk6          1  
        cn_ss6          1  
        cn_prsity6      1  
        welflux         2  
        welflux_k02      6  
        drncond_k00     10  
        pp_hk2          32  
        pp_rech1        32  
        pp_sy1          32  
        pp_vka2          32  
        pp_rech0        32  
        pp_hk1          32  
        pp_prsity2      32  
        pp_prsity0      32  
        pp_ss2          32  
        pp_sy2          32  
        pp_hk0          32  
        pp_ss1          32  
        pp_vka0          32  
        pp_strt1        32
```

```

pp_vka1      32
pp_ss0       32
pp_strt0     32
pp_strt2     32
pp_sy0       32
pp_prsity1   32
strk         40
Name: pargp, dtype: int64

```

fix the future recharge pilot points, vka in layers 1 and 3 and the initial condition pilot points (we still have layer-scale pars for each of these types)

```

In [10]: fi_grps = ["pp_rech1", "pp_vka0", "pp_vka2", "pp_strt0", "pp_strt1", "pp_strt2"]
         par.loc[par.pargp.apply(lambda x: x in fi_grps), "partrans"] = "fixed"
         pst.npar_adj

```

```

Out[10]: 527

```

Ok, thats better...so lets run PESTPP-GLM. We will use a single "base parameter" jacobian matrix as the basis for 6 super parameter iterations. Then we will draw 100 realizations from the FOSM posterior parameter covariance matrix and run those 100 realizations to get the psoterior forecast PDFs

```

In [11]: pst.control_data.noptmax = 3
         pst.pestpp_options["n_iter_base"] = -1
         pst.pestpp_options["n_iter_super"] = 3
         pst.pestpp_options["num_reals"] = 50 # this is how many ies uses
         pst.pestpp_options["parcov"] = "prior_cov.jcb"
         pst.write(os.path.join(t_d, "freyberg_pp.pst"))

```

```

In [12]: pyemu.os_utils.start_slaves(t_d, "pestpp-glm", "freyberg_pp.pst", num_slaves=20, slave_ro
         master_dir=m_d)

```

```

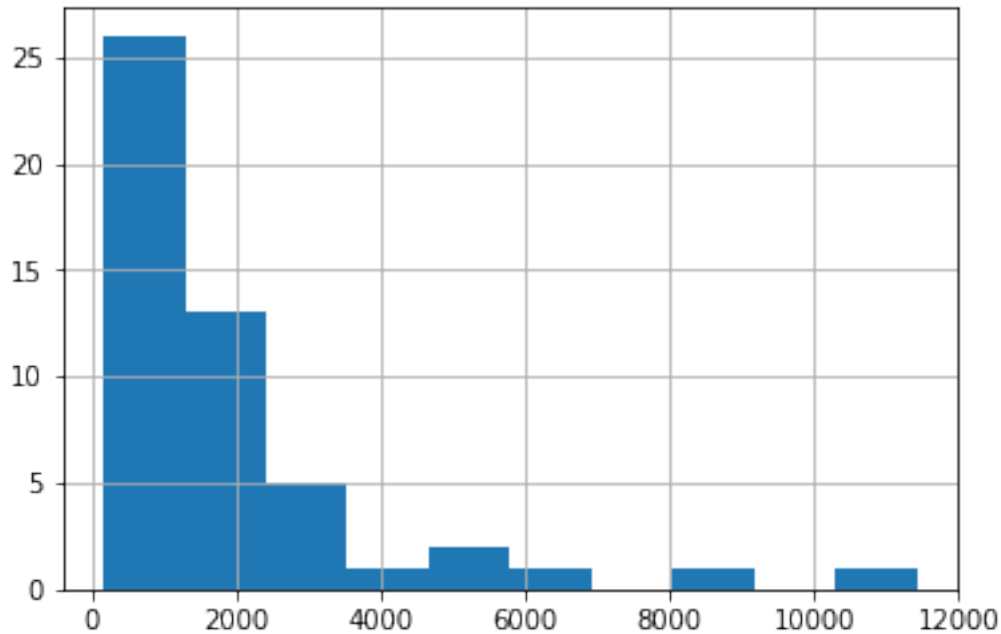
In [13]: df = df=pd.read_csv(os.path.join(m_d, "freyberg_pp.post.obsen.csv"), index_col=0)
         oe = pyemu.ObservationEnsemble.from_dataframe(pst=pst, df=df)

```

```

In [14]: ax = oe.phi_vector.hist() #bins=np.linspace(0, 100, 20))

```



Here we see the distribution of phi values across the 100 posterior realizations. Should we accept all of these??? The theoretical phi we should accept is number of nonzero obs (14).

To get a “posterior” ensemble, we need to throw out the realizations with large phi - lets just take the 20 best:

```
In [15]: oe_pt = oe.loc[oe.phi_vector.sort_values().index[:20],:] #just take the 20 lowest phi
```

We can also load and plot the FOSM forecast results along side of the ensemble results:

```
In [16]: f_df = pd.read_csv(os.path.join(m_d,"freyberg_pp.pred.usum.csv"),index_col=0)
f_df.index = f_df.index.map(str.lower)
f_df
```

```
Out[16]:
```

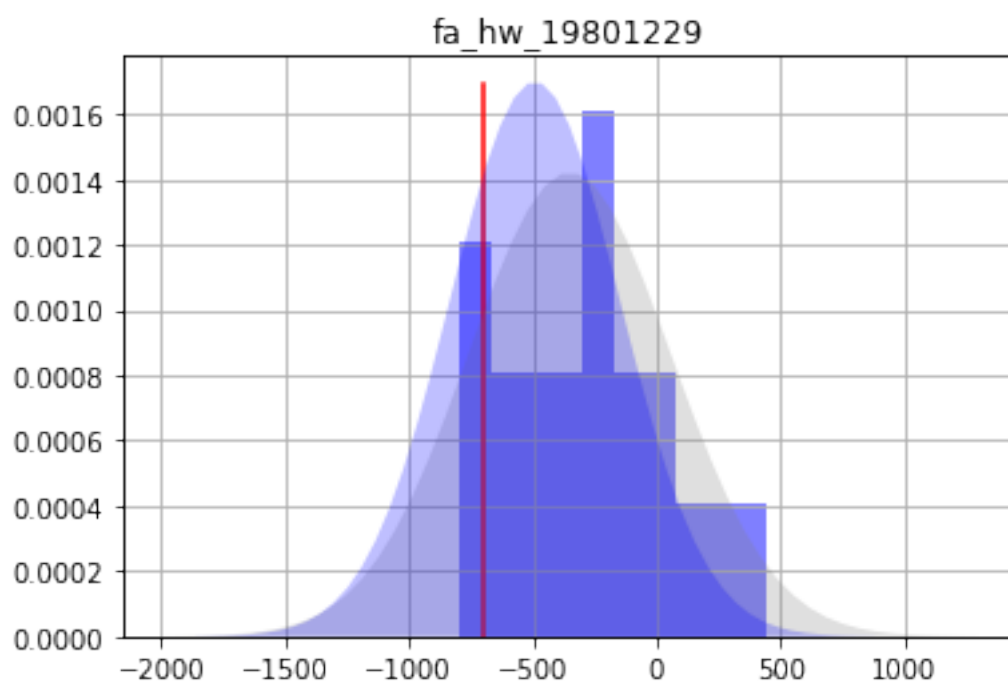
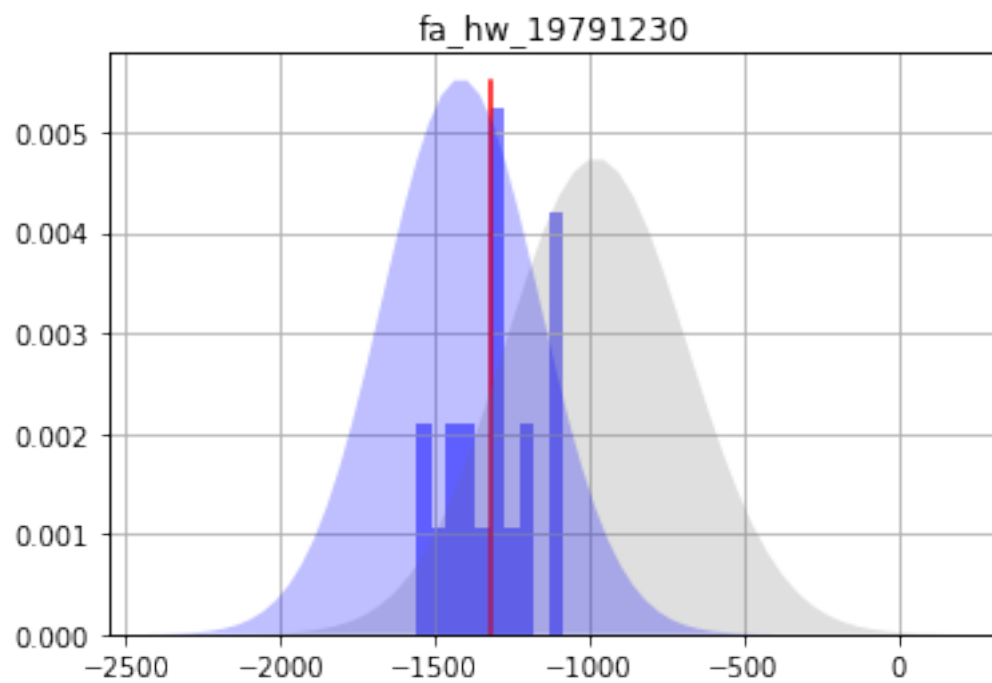
	prior_mean	prior_stdev	prior_lower_bound \
name			
fa_hw_19791230	-977.2390	295.32800	-1567.8900
fa_hw_19801229	-351.2160	409.77000	-1170.7600
fa_tw_19791230	-453.0330	409.35100	-1271.7400
fa_tw_19801229	108.9600	506.73200	-904.5040
hds_00_013_002_000	39.6102	3.96314	31.6840
hds_00_013_002_001	38.3838	4.05782	30.2681
part_status	2.0000	0.00000	2.0000
part_time	907.7020	704.75100	-501.8010

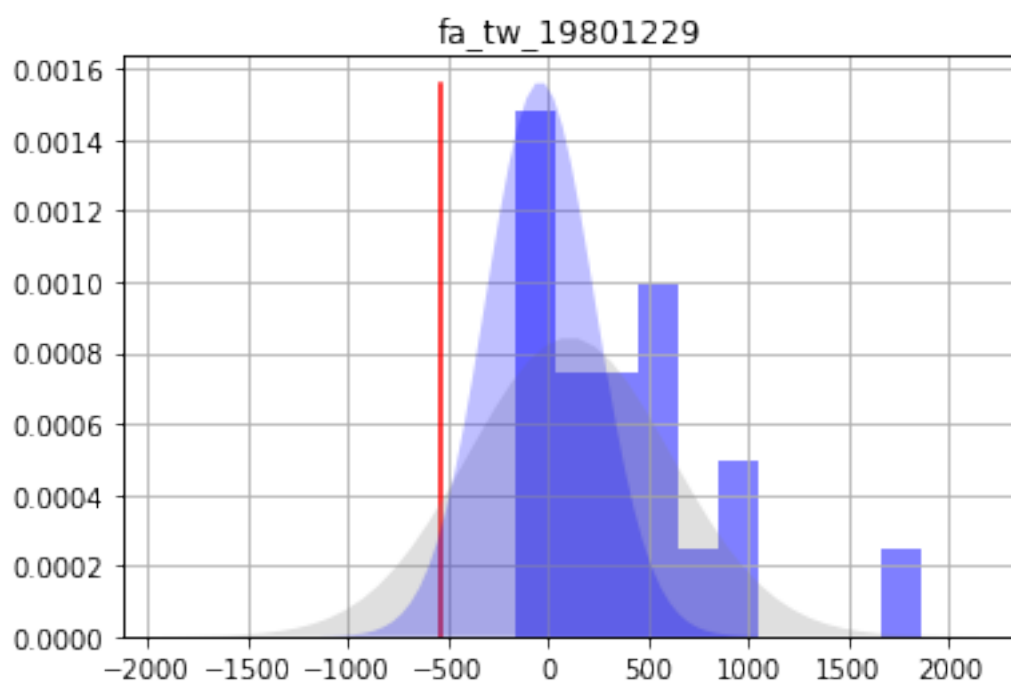
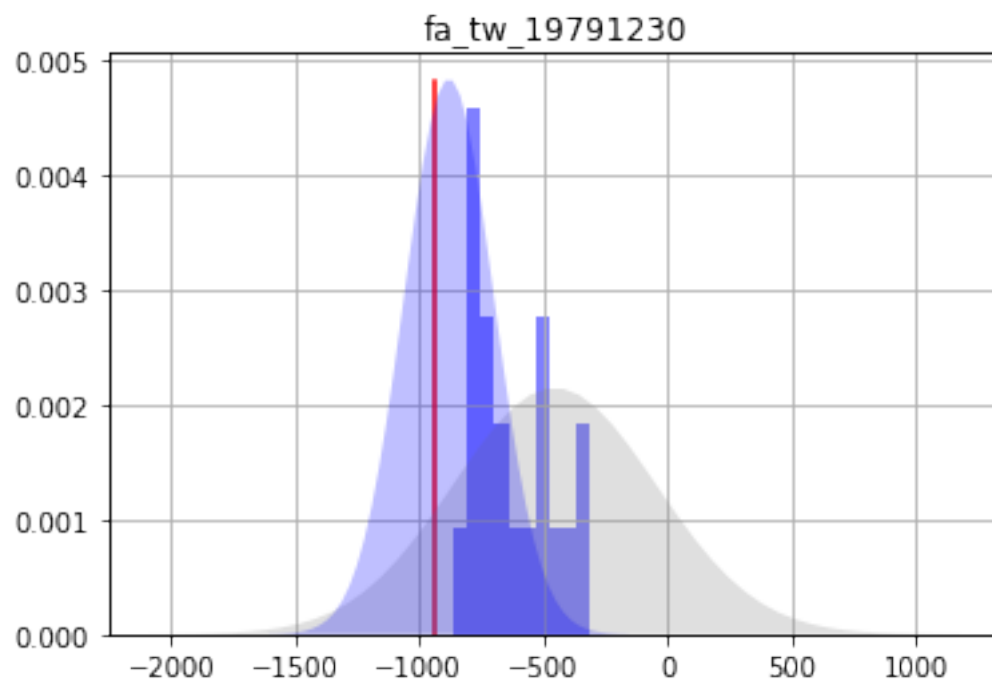
	prior_upper_bound	post_mean	post_stdev \
name			
fa_hw_19791230	-386.5840	-1415.6100	252.974000

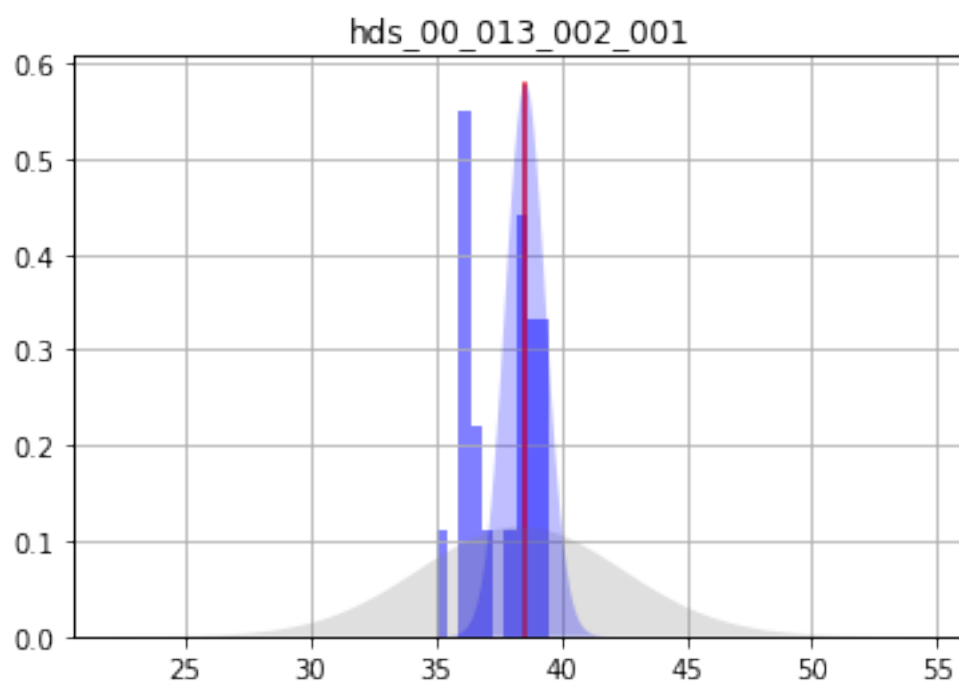
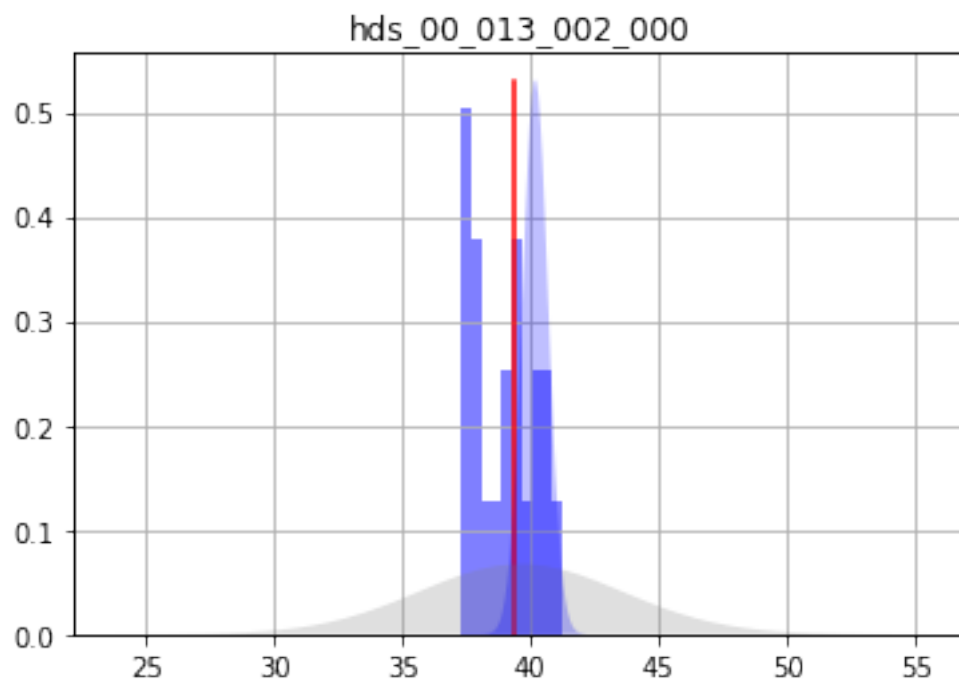
fa_hw_19801229	468.3240	-491.6660	342.574000
fa_tw_19791230	365.6690	-878.9190	181.462000
fa_tw_19801229	1122.4200	-38.3736	272.738000
hds_00_013_002_000	47.5365	40.1763	0.501236
hds_00_013_002_001	46.4994	38.5735	0.802533
part_status	2.0000	2.0000	0.000000
part_time	2317.2000	916.9650	606.568000

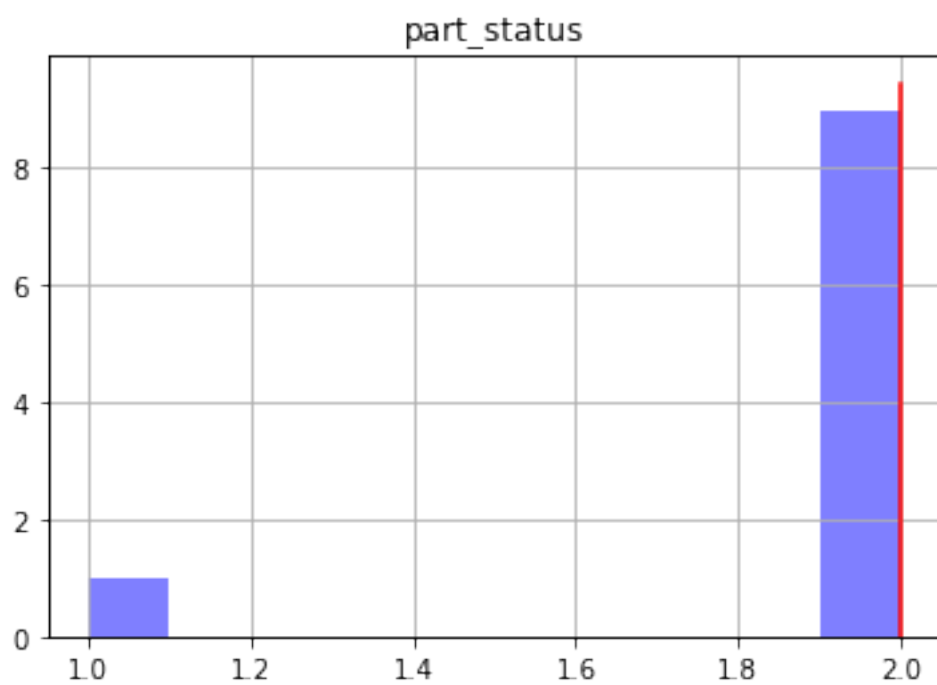
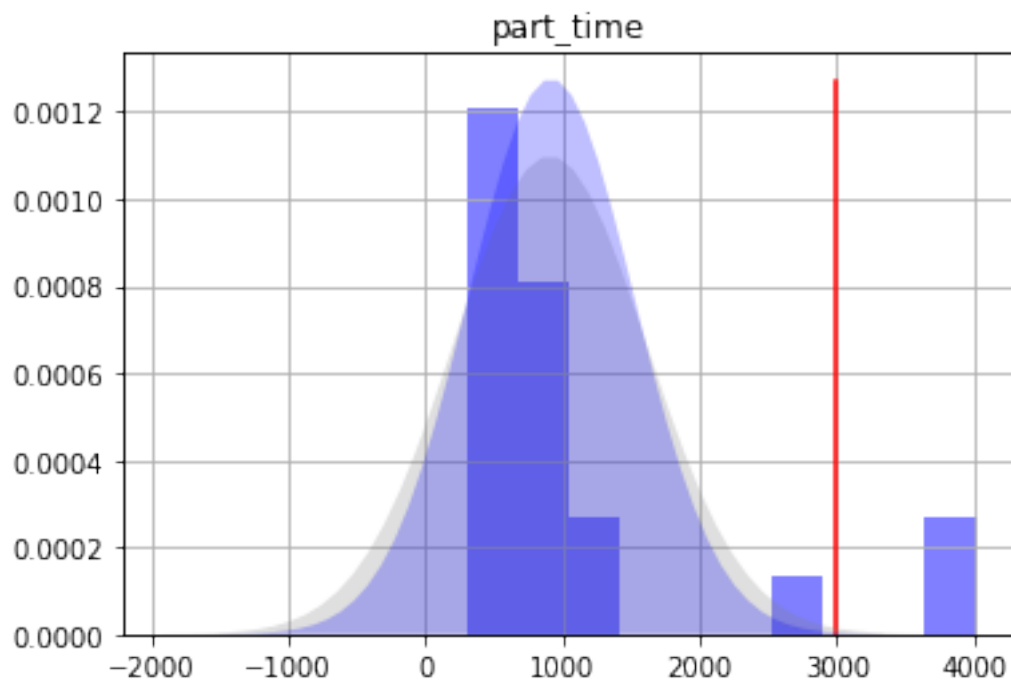
	post_lower_bound	post_upper_bound
name		
fa_hw_19791230	-1921.5600	-909.6600
fa_hw_19801229	-1176.8100	193.4820
fa_tw_19791230	-1241.8400	-515.9940
fa_tw_19801229	-583.8500	507.1030
hds_00_013_002_000	39.1738	41.1787
hds_00_013_002_001	36.9685	40.1786
part_status	2.0000	2.0000
part_time	-296.1710	2130.1000

```
In [17]: obs = pst.observation_data
fnames = pst.pestpp_options["forecasts"].split(",")
for forecast in fnames:
    ax = plt.subplot(111)
    oe_pt.loc[:,forecast].hist(ax=ax,color="b",alpha=0.5,normed=True)
    ax.plot([obs.loc[forecast,"obsval"],obs.loc[forecast,"obsval"]],ax.get_ylim(),"r")
    axt = plt.twinx()
    x,y = pyemu.plot_utils.gaussian_distribution(f_df.loc[forecast,"prior_mean"],f_df.loc[forecast,"prior_std"])
    axt.fill_between(x,0,y,facecolor="0.5",alpha=0.25)
    x,y = pyemu.plot_utils.gaussian_distribution(f_df.loc[forecast,"post_mean"],f_df.loc[forecast,"post_std"])
    axt.fill_between(x,0,y,facecolor="b",alpha=0.25)
    axt.set_ylim(0,axt.get_ylim()[1])
    axt.set_yticks([])
    ax.set_title(forecast)
plt.show()
```







1.0.2 Setup of Tikhonov regularization

Now lets setup and use some formal regularization to bring the final phi up to around 14. We will use first-order regularization based on the covariance matrix we build earlier:

```
In [18]: cov = pyemu.Cov.from_binary(os.path.join(t_d,"prior_cov.jcb"))
```

```
new binary format detected...
```

```
In [19]: cnames = set(cov.row_names)
         pnames = set(pst.adj_par_names)
         cnames.symmetric_difference(pnames)
```

```
Out[19]: {'ss3021007',
          'sy4017000',
          'ss5008013',
          'sy5028013',
          'rech3012017',
          'prsity5026003',
          'vka4035017',
          'hk4029016',
          'strt4007006',
          'ss4007003',
          'vka3039013',
          'vka4009018',
          'vka3030005',
          'sy4004011',
          'sy4014014',
          'strt5018010',
          'rech2010011',
          'vka229',
          'vka5026016',
          'ss5004013',
          'sy4011018',
          'hk5014017',
          'hk4009013',
          'strt4011008',
          'prsity3009015',
          'vka3002010',
          'vka3021017',
          'prsity5019018',
          'ss5007008',
          'ss3026012',
          'strt4028003',
          'vka3035006',
          'vka5002001',
          'ss5030012',
          'ss5006008',
```

'hk3014015',
'vka4035007',
'prsity3015019',
'hk4018008',
'rech3014002',
'sy3030008',
'hk3026004',
'prsity4038010',
'vka3030013',
'strt5003011',
'hk4000005',
'strt3038011',
'prsity4009016',
'prsity3014015',
'ss4011015',
'hk3021001',
'prsity4039008',
'strt4020014',
'rech2010008',
'rech2032006',
'hk4035014',
'strt5031011',
'prsity5011019',
'vka3035005',
'hk3011008',
'prsity4025015',
'sy3018015',
'prsity4035005',
'sy5016009',
'sy3024015',
'prsity3020012',
'hk4002010',
'rech2000013',
'hk4027010',
'ss5003019',
'strt5036012',
'hk4004018',
'ss4003010',
'ss4010011',
'prsity3035016',
'rech2021018',
'prsity5036005',
'ss5026009',
'hk4031004',
'rech3000001',
'ss4005008',
'rech3017008',
'ss4033011',

'strt4016018',
'strt5030006',
'sy5015008',
'strt4007005',
'prsity4028017',
'sy4020001',
'hk3014009',
'strt3027002',
'hk5030007',
'prsity3029016',
'sy4029017',
'vka5028016',
'strt3014001',
'strt3037013',
'sy4019011',
'sy4024002',
'vka4011018',
'vka228',
'strt5036006',
'prsity3002000',
'sy5033009',
'vka5037008',
'hk3031004',
'strt3036017',
'prsity3039008',
'sy3000004',
'prsity4025009',
'hk4005013',
'strt4002007',
'rech2037013',
'sy5021017',
'rech3013018',
'sy4002004',
'hk3037014',
'prsity4037005',
'ss3016017',
'rech3028010',
'vka3003019',
'prsity3024014',
'strt3032009',
'hk3008018',
'sy3028011',
'rech2014001',
'rech2022007',
'sy4009013',
'vka5034011',
'hk5001014',
'vka5007004',

'ss4019012',
'strt4034006',
'rech3005005',
'sy3012014',
'ss3003012',
'sy3022013',
'strt4039014',
'vka5005000',
'vka3025011',
'rech3024016',
'ss3035015',
'prsity4036012',
'ss3033009',
'hk4037014',
'hk5034010',
'strt4038012',
'ss3022008',
'sy3009001',
'ss3031019',
'vka4026016',
'vka4004004',
'hk3000011',
'sy5003011',
'ss5031017',
'vka3003006',
'rech2017000',
'rech3035013',
'sy5021001',
'rech3016017',
'strt5000013',
'prsity4039009',
'rech2028007',
'hk4002013',
'rech3014018',
'vka5037006',
'vka4017019',
'sy3025002',
'sy3013010',
'strt3002012',
'hk5026011',
'ss4025010',
'prsity4015015',
'strt4030009',
'strt3024013',
'hk5023016',
'strt225',
'prsity5006005',
'prsity4003012',

'hk4017010',
'hk4025001',
'hk3038008',
'prsity3035013',
'ss5031001',
'strt4000002',
'sy3001014',
'prsity3010010',
'sy3028002',
'sy5000018',
'rech3033007',
'prsity3019002',
'vka3000015',
'strt5025006',
'rech2025013',
'ss5002016',
'sy5008002',
'vka5010014',
'strt5033007',
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```

```
In [20]: pyemu.helpers.first_order_pearson_tikhonov(pst,cov)
```

```

getting CC matrix
processing

```

```
In [21]: pst.prior_information.head()
```

```

Out[21]:

```

	equation	obgnme	\
pilbl			
pcc_1	1.0 * log(dc0000390005) - 1.0 * log(dc0000390006) = 0.0	regul_cc	
pcc_2	1.0 * log(dc0000390005) - 1.0 * log(dc0000390007) = 0.0	regul_cc	
pcc_3	1.0 * log(dc0000390005) - 1.0 * log(dc0000390008) = 0.0	regul_cc	
pcc_4	1.0 * log(dc0000390005) - 1.0 * log(dc0000390009) = 0.0	regul_cc	
pcc_5	1.0 * log(dc0000390005) - 1.0 * log(dc0000390010) = 0.0	regul_cc	

	pilbl	weight
pilbl		
pcc_1	pcc_1	0.904837
pcc_2	pcc_2	0.818731
pcc_3	pcc_3	0.740818
pcc_4	pcc_4	0.670320
pcc_5	pcc_5	0.606531

```
In [22]: shutil.copy2(os.path.join(m_d,"freyberg_pp.jcb"),os.path.join(t_d,"restart_pp.jcb"))
```

```
Out[22]: 'template/restart_pp.jcb'
```

```

In [23]: pst.pestpp_options["base_jacobian"] = "restart_pp.jcb"
pst.reg_data.phimlim = pst.nnz_obs
pst.reg_data.phimaccept = pst.reg_data.phimlim * 1.1
pst.write(os.path.join(t_d,"freyberg_pp.pst"))

```

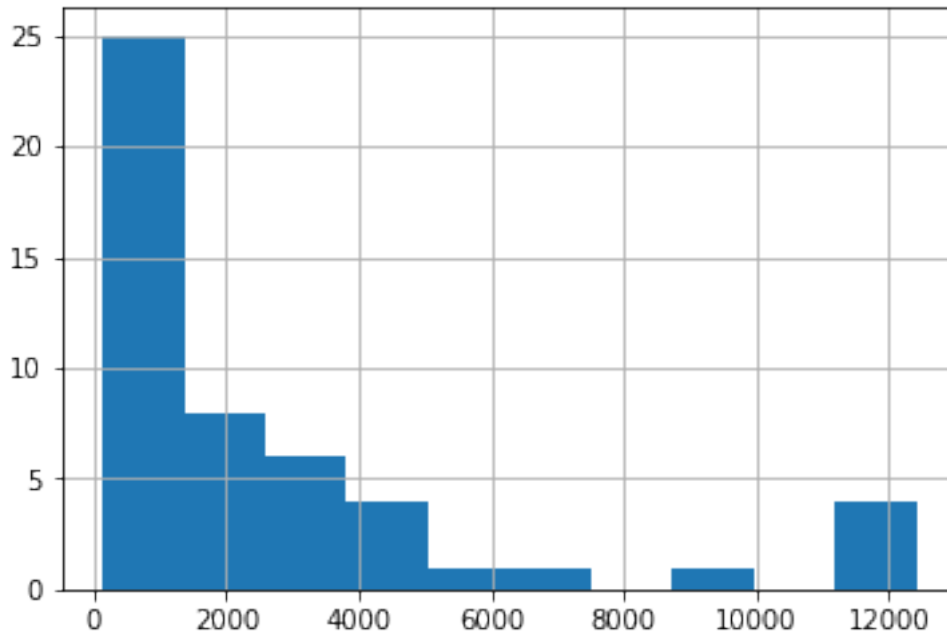
```
In [24]: pyemu.os_utils.start_slaves(t_d,"pestpp-glm","freyberg_pp.pst",num_slaves=20,slave_ro
master_dir=m_d)
```

```

In [25]: df = df=pd.read_csv(os.path.join(m_d,"freyberg_pp.post.obsen.csv"),index_col=0)
oe = pyemu.ObservationEnsemble.from_dataframe(pst=pst,df=df)

```

```
In [26]: ax = oe.phi_vector.hist()#bins=np.linspace(0,100,20))
```



Same as before, to get a “posterior” ensemble, we need to throw out the realizations with large phi - lets just take the 20 best:

```
In [27]: oe_pt = oe.loc[oe.phi_vector.sort_values().index[:20],:]
```

```
In [28]: f_df = pd.read_csv(os.path.join(m_d,"freyberg_pp.pred.usum.csv"),index_col=0)
f_df.index = f_df.index.map(str.lower)
f_df
```

```
Out[28]:
```

	prior_mean	prior_stdev	prior_lower_bound	\
name				
fa_hw_19791230	-977.2390	295.32800	-1567.8900	
fa_hw_19801229	-351.2160	409.77000	-1170.7600	
fa_tw_19791230	-453.0330	409.35100	-1271.7400	
fa_tw_19801229	108.9600	506.73200	-904.5040	
hds_00_013_002_000	39.6102	3.96314	31.6840	
hds_00_013_002_001	38.3838	4.05782	30.2681	
part_status	2.0000	0.00000	2.0000	
part_time	907.7020	704.75100	-501.8010	

	prior_upper_bound	post_mean	post_stdev	\
name				
fa_hw_19791230	-386.5840	-1226.1600	253.662000	
fa_hw_19801229	468.3240	-438.0650	342.835000	
fa_tw_19791230	365.6690	-685.1120	185.061000	
fa_tw_19801229	1122.4200	25.8356	275.087000	
hds_00_013_002_000	47.5365	40.1357	0.528593	

hds_00_013_002_001	46.4994	38.6689	0.821579
part_status	2.0000	2.0000	0.000000
part_time	2317.2000	864.5700	609.653000

	post_lower_bound	post_upper_bound
name		
fa_hw_19791230	-1733.4900	-718.8380
fa_hw_19801229	-1123.7400	247.6060
fa_tw_19791230	-1055.2300	-314.9900
fa_tw_19801229	-524.3380	576.0100
hds_00_013_002_000	39.0785	41.1929
hds_00_013_002_001	37.0257	40.3120
part_status	2.0000	2.0000
part_time	-354.7370	2083.8800

```
In [29]: obs = pst.observation_data
fnames = pst.pestpp_options["forecasts"].split(",")
for forecast in fnames:
    ax = plt.subplot(111)
    oe_pt.loc[:,forecast].hist(ax=ax,color="b",alpha=0.5,normed=True)
    ax.plot([obs.loc[forecast,"obsval"],obs.loc[forecast,"obsval"]],ax.get_ylim(),"r")
    axt = plt.twinx()
    x,y = pyemu.plot_utils.gaussian_distribution(f_df.loc[forecast,"prior_mean"],f_df.loc[forecast,"prior_std"])
    axt.fill_between(x,0,y,facecolor="0.5",alpha=0.25)
    x,y = pyemu.plot_utils.gaussian_distribution(f_df.loc[forecast,"post_mean"],f_df.loc[forecast,"post_std"])
    axt.fill_between(x,0,y,facecolor="b",alpha=0.25)
    axt.set_ylim(0,axt.get_ylim()[1])
    axt.set_yticks([])
    ax.set_title(forecast)
plt.show()
```

