University of Cincinnati

GEOL 6024: GroundWater Modeling

Water Withdrawal Controversy

Project Report

Submitted By

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Contents

1. Introduction

Prairie Valley Water is installing a high-capacity well to support the expanded bottled water operations. But the company's permit is denied by the state because the state's water withdrawal assessment tool predicted that their proposed well could adversely impact the nearby trout stream. The company argues that their well will pump deep groundwater which will have no impact on what happens at the surface.
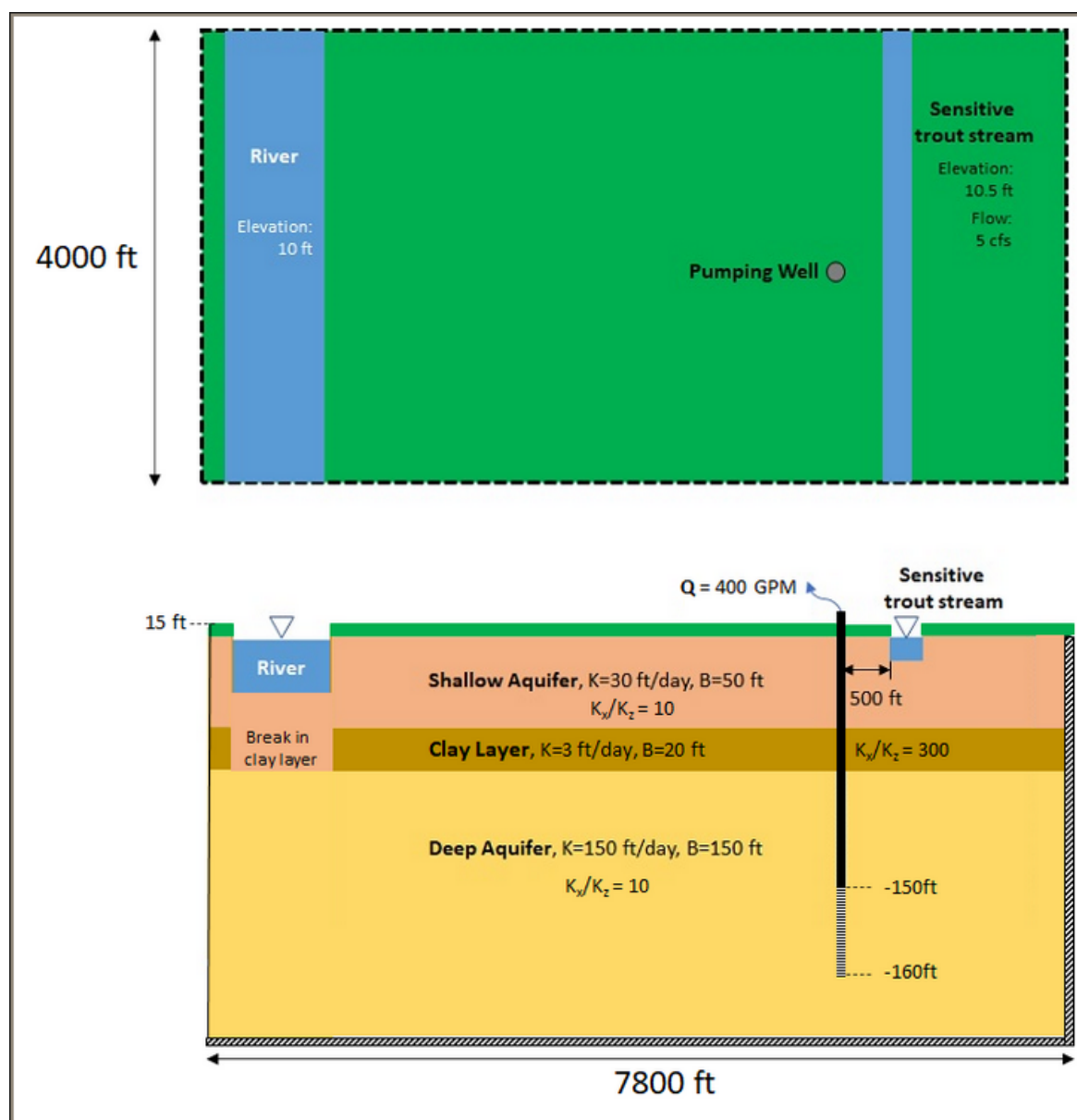
Here, we're going to make a groundwater model based on the given situation. Since we can't know for sure how much water will flow in the river and the stream as they have upstream contributions, we'll look at the rainwater contributions to the stream from the given area, and compare how much rainwater recharge to the stream when there is well compared to when there was no well.

And if the recharge to the stream is less than 5\

2. Code

2.1 Importing libraries

Importing the libraries. Here `geometry` package from `shapely` library is being used for the defining the geometries of rivers as polygons so we can easily get the grid points that are inside it.

--- code: python ---

```python
import flopy
import matplotlib.pyplot as plt
import numpy as np
```

```python
4
5    from shapely import geometry
```

2.2 Simulation Parameters

Parameters for the grids.

```python
1    X0 = 0
2    XN = 7800
3    NC = 78
4    ΔX = 7800/NC
5
6    Y0 = 0
7    YN = 4000
8    NR = 40
9    ΔY = 4000/NR
10
11
12   Top = 15
13   Height = 220
14   Bottom = Top-Height
```

Geo Layers, these are geologically different layers, the subdivisions contains the number of divisions each of these layers will contain, and the total of that will be the actual layers used by flopy.

```python
1    geolyr_thickness = [50, 20, 150]
2    geolyr_subdivisions = [10, 4, 30]
```

This variable will be used and if it's true only then the well will be added to the model.

```python
1    WELL_ON = True
```

Here, 2D grid is defined using numpy, we can use this for each layer and to have other informations like distances.

```python
1    xy_grid_points = np.mgrid[X0:XN:ΔX, Y0:YN:ΔY].reshape(2, -1).T
2    x_grids = np.linspace(X0, XN, NC)
```

Here we use the geometry package to define the geometries of different zones, and well as well have any other relevant parameters that'll be used by the model later.

```python
domain = geometry.box(X0, Y0, XN, YN)
river = geometry.box(300, 0, 300+850, YN)
river_top = 10
river_height = 25
river_bottom = river_top - river_height
river_conductance = 50 * ΔX * ΔY  # leakance to conductance
stream = geometry.box(6150, 0, 6150+270, YN)
stream_top = 10.5
stream_height = 3.3
stream_conductance = 1 * ΔX * ΔY  # leakance to conductance
stream_bottom = stream_top - stream_height
well = geometry.Point((5720, 2000))
well_top = -150
well_bottom = -160
well_rate = -200 * 0.1336801*60*24  # GPM → ft³/day
```

2.3 Utility Functions and Pre-processing

Let's define a function that gives you the layers it includes if you give it the top and bottom elevation of something.

For example, if you need the layers for well, you can call this function with well screening top and bottom elevation.

This will return a generator for tuples containing the layer index, top elevation and bottom elevation. The elevations are important in case of partial covering of the layer.

```python
def get_layers(top=Top, bottom=Bottom):
    all_layers = [(i, b) for i, b in enumerate(bot) if b < top]
    for i, b in all_layers:
        if b > bottom:
            yield i, top, b
        else:
            break
        top = b
    if b <= bottom:
        yield i, top, bottom
```

Similarly, a little more complex function to get the grid points. You can provide a shape and it'll give you the grid points, to be safe use `geometry.Polygon` shape. But you can use any shape if it has `.contains()` method.

It'll return a generator for tuples containing (lay, row, col). if layers is not given it'll return points on top most layer (index=0).

```python
def get_grid_points(shape, layers=None):
    if not layers:
```

```python
3          layers = [0]
4      else:
5          layers = list(layers)
6      for i, gp in enumerate(xy_grid_points):
7          col = i // (NR)              # might have to swap these two.
8          row = i % (NR)
9          pt = geometry.Point(gp[0], gp[1])
10         if shape.contains(pt):
11             # layer, row, col
12             for j in layers:
13                 yield (j, row, col)
```

Now we make the actual computational layers, also define a lookup array that relates computational layers to the geolayers they are part of.

`lookup_table[i]` should give `j`, where `i` is the index of computational layer and `j` is the index of geo layer.

———————————————————— code: python ————————————————————
```python
1  # computational layers
2  NLay = sum(geolyr_subdivisions)
3  lookup_table = np.concatenate(
4      list(np.ones(s, dtype=int)*i for i, s in
5          enumerate(geolyr_subdivisions)))
```

For **implementing hetereogeiniety** in second geolayer (index=1), we have to have different values of K in the same layer.

As we know the break in clay layer is in the same area as the river, so we'll use the grid points belonging to the river, and modify the K value of the second layer for those grid points only.

———————————————————— code: python ————————————————————
```python
1  # hetereogeiniety in 2nd geolayer
2  k_2nd_layer = np.ones(shape=(NR, NC))*3.0
3  kv_2nd_layer = np.ones(shape=(NR, NC))*.01
4  for cell in get_grid_points(river):
5      k_2nd_layer[cell[1], cell[2]] = 30.0
6      kv_2nd_layer[cell[1], cell[2]] = 3.0
```

Now the overall K values for 3 geolayers will be something like this.

———————————————————— code: python ————————————————————
```python
1  lyr_k_hz = [30.0,
2              k_2nd_layer,
3              150.0]
4  lyr_k_vt = [3.0,
5              kv_2nd_layer,
6              15.0]
```

Now to generate the data for each computational layers.

```python
thickness = np.zeros(NLay)
k_hz = [0 for i in range(NLay)]
k_vt = [0 for i in range(NLay)]
bot = np.ones(NLay)
```

This will loop through the computational layers and then will use the lookup table to fill the values from geolayers.

```python
for lay in range(NLay):
    geo_lay = lookup_table[lay]
    thickness[lay] =
    ↪ geolyr_thickness[geo_lay]/geolyr_subdivisions[geo_lay]
    k_hz[lay] = lyr_k_hz[geo_lay]
    k_vt[lay] = lyr_k_vt[geo_lay]
    bot[lay] = Top-sum(thickness)
```

Now for riv package we need to give stress period, so let's define a function to generate the stress period data.

From the documentation we can see we need to return a tuple with (cellid, stage, cond, rbot, aux, boundname), with last 2 being optional.

We'll use the layers falling between the river$_\text{top}$ and bottom elevation, and then use the shape to get the grid points, for a simple model we'll just use the same conductance value for each cell. We can modify that using how much of the river the cell contains but here rivers are polygons so let's ignore that for now.

We'll do the same for river and the stream and return a generator with stress period data.

```python
def get_riv_stress_period():
    "gives the stress_period_data on the grid_points for river
    ↪ grids."

    layers_tuple = list(get_layers(top=river_top,
    ↪ bottom=river_bottom))
    for grid_pt in get_grid_points(river):
        for lay, thk, bottom in layers_tuple:
            # cellid, stage, cond, rbot, aux, boundname
            yield ((lay, grid_pt[1], grid_pt[2]),
                   thk, river_conductance, bottom)
    layers_tuple = list(get_layers(top=stream_top,
    ↪ bottom=stream_bottom))
    for grid_pt in get_grid_points(stream):
        for lay, thk, bottom in layers_tuple:
```

```python
13              yield ((lay, grid_pt[1], grid_pt[2]),
14                      thk, stream_conductance, bottom)
```

Similarly, we define the stress period data for chd package, which is for constant head boundaries. We'll use the same concept as before but return a generator for a tuple containing (cellid, head) value.

```python
# code: python
1  def get_chd_stress_period():
2      "gives the stress_period_data on the grid_points for constant
       ↪  head points."
3      # river grid points
4      layers_tuple = list(get_layers(top=river_top,
       ↪  bottom=river_bottom))
5      for grid_pt in get_grid_points(river):
6          for lay, thk, bottom in layers_tuple:
7              # cellid, head
8              yield ((lay, grid_pt[1], grid_pt[2]), 10)
9
10     # stream grid points
11     layers_tuple = list(get_layers(top=stream_top,
       ↪  bottom=stream_bottom))
12     for grid_pt in get_grid_points(stream):
13         for lay, thk, bottom in layers_tuple:
14             yield ((lay, grid_pt[1], grid_pt[2]), 10.5)
```

Similarly, function for well stress period. Here we're trying to find the grid point closest to the well co-ordinates with same logic as in the get_grid_points function but for a point.

```python
# code: python
1  _gps = map(geometry.Point, xy_grid_points)
2  _well_gp = min(enumerate(_gps), key=lambda x: well.distance(x[1]))
3  well_row = _well_gp[0] % (NR)
4  well_col = _well_gp[0] // (NR)
5
6  def get_well_stress_period():
7      well_layers = list(get_layers(well_top, well_bottom))
8      return {0: [((i, well_row, well_col),
9                  well_rate/len(well_layers)) for i, _, _ in
10                 well_layers]}
```

To make sure our function is working as intended, we can call it once and plot that data. Let's plot the chd stress period data with colormap of heads.

```python
# code: python
1  sp = list(get_chd_stress_period())
2
3  ipoints = np.ones((NLay, NR, NC))
```

```python
4  for i, _ in sp:
5      ipoints[i] = -1
6  x = [l[0][2] for l in sp]
7  y = [l[0][1] for l in sp]
8  c = [l[1] for l in sp]
```
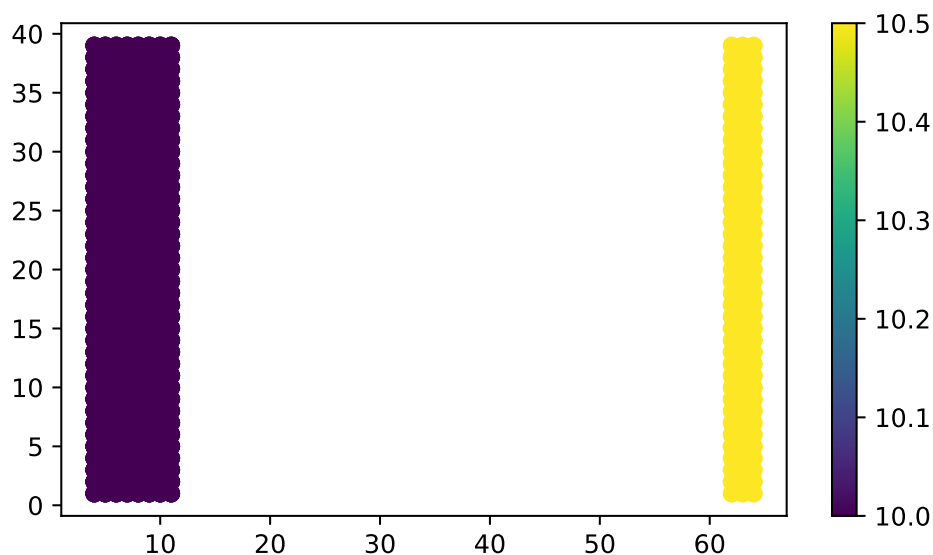
After we have the data, we can plot it.

```python
code: python
1  plt.scatter(x, y, c=c)
2  plt.colorbar()
3  plt.show()
```


The plot shows two patches of constant head boundaries where our river and trout streams are so it looks good.

2.4 Flopy Model

Now we're ready to build the model. First things are to define the path and model name, as well as link the modflow executable.

```python
code: python
1  ws = './models/3_water_withdrawal_controversy'
2  name = '3_water_wd'
3
4  sim = flopy.mf6.MFSimulation(sim_name=name,
5                               sim_ws=ws,
6                               exe_name='modflow-mf6')
```

Here we'll define `tdis`, `ims` and `gwf` packages.

```python
tdis = flopy.mf6.ModflowTdis(sim,
                             time_units='days')
ims = flopy.mf6.ModflowIms(sim)
gwf = flopy.mf6.ModflowGwf(sim, modelname=name, save_flows=True)
```

Now the discretization (`dis`) package will use the grid information to generate a grid representation.

```python
dis = flopy.mf6.ModflowGwfdis(gwf,
                              length_units='FEET',
                              nlay=NLay,
                              nrow=NR,
                              ncol=NC,
                              delc=ΔX,
                              delr=ΔY,
                              top=Top,
                              botm=bot)
```

We'll use the top elevation as initial head, and pass that to `ic` package.

```python
initial_head = np.ones((NLay, NR, NC)) * Top
ic = flopy.mf6.ModflowGwfic(gwf, strt=initial_head)
```

We'll use 1ft/year of recharge, and convert it to ft/day as our units for length and time are in feet and day respectively.

We'll also define the rivers and the conductivity with hetereogeiniety here.

```python
recharge = flopy.mf6.ModflowGwfrcha(gwf, recharge=1/365)
rivers = flopy.mf6.ModflowGwfriv(
    gwf,
    stress_period_data=list(get_riv_stress_period()))
npf = flopy.mf6.ModflowGwfnpf(gwf,
                              icelltype=1,
                              k=k_hz,
                              k33=k_vt,
                              save_specific_discharge=True)
```

This commented out code here shows how we can modify the values of k even after we have made the model, and we can set the data of model with out modified data.

We can use this for making that hetereogeiniety if we don't want to do it at the beginning.

---- code: python ----

```python
# EXample to modify the k values after it is defined.
# k_values = npf.k.get_data()
# kv_values = npf.k33.get_data()
# layers_2nd = [i for i, v in enumerate(lookup_table) if v == 1]
# for lay in layers_2nd:
#     k_values[lay] = k_2nd_layer
#     kv_values[lay] = kv_2nd_layer
# npf.k.set_data(k_values)
# npf.k33.set_data(kv_values)
```

Now we'll use the previously defined functions to define the chd package.

---- code: python ----

```python
chd = flopy.mf6.ModflowGwfchd(
    gwf,
    stress_period_data=list(get_chd_stress_period()))
```

If well is to be inserted, insert it.

---- code: python ----

```python
if WELL_ON:
    wells = flopy.mf6.ModflowGwfwel(
        gwf,
        stress_period_data=get_well_stress_period())
```

Now we'll define some files to write the output of simulation too.

---- code: python ----

```python
budget_file = name + '.bud'
head_file = name + '.hds'
oc = flopy.mf6.ModflowGwfoc(gwf,
                            budget_filerecord=budget_file,
                            head_filerecord=head_file,
                            saverecord=[('HEAD', 'ALL'),
                                        ('BUDGET', 'ALL')])
```

Finally, write all those details to modflow input files, and then run it.

---- code: python ----

```python
sim.write_simulation()
result, _ = sim.run_simulation()

if not result:
    print("Error in Simulation")
    exit(1)
```

If we encounter errors we'd have to check what we made mistakes on and repeat it. Since it's good for now, let's try to extract the outputs.
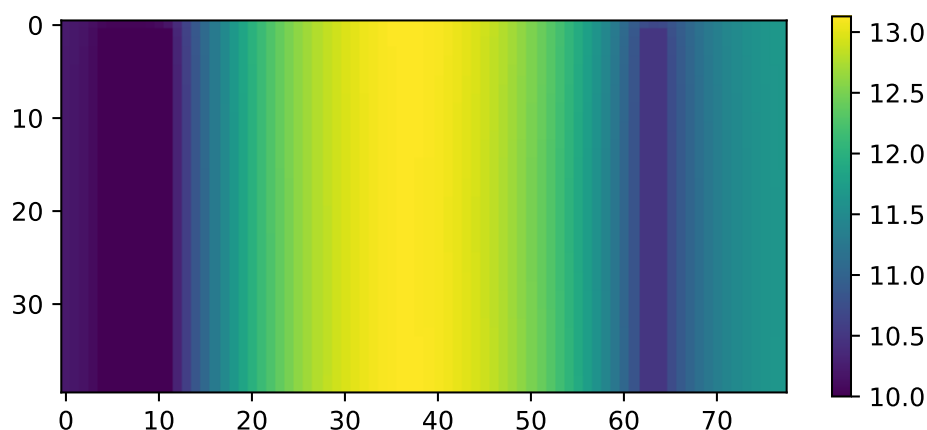
```python
head_arr = gwf.output.head().get_data()
bud = gwf.output.budget()

chd_bud = bud.get_data(text='CHD')


spdis = bud.get_data(text='DATA-SPDIS')[0]
qx, qy, qz =
   flopy.utils.postprocessing.get_specific_discharge(spdis, gwf)
watertable = flopy.utils.postprocessing.get_water_table(head_arr,
   -1e30)
```

We used some postprocessing tools to get the outputs we wanted, let's plot it and see.

```python
plt.imshow(watertable)
plt.colorbar()
plt.show()
```


We can see the river heads are ok, and then the heads decrease fro the middle of the simulation towards the river streams.

```python
def plot_plan(ext='pdf', layer=0):
    fig, ax = plt.subplots(1, 1, figsize=(9, 3),
       constrained_layout=True)
    ax.set_title(f'Layer-{layer}')
    pmv = flopy.plot.PlotMapView(gwf, ax=ax)
    pmv.plot_array(head_arr[layer])
    pmv.plot_grid(colors='white', linewidths=0.3)
    pmv.contour_array(head_arr[layer],
```

```python
8                      linewidths=1.,
9                      cmap='Wistia')
10      # flopy.plot.styles.graph_legend()
11      pmv.plot_vector(qx[layer, :, :], qy[layer, :, :],
12                      normalize=False, istep=2, jstep=2,
                        ↪ color="white")
13      filename = f"./images/3_plan_layer-{layer}.{ext}"
14      plt.savefig(filename)
15      plt.show()
16      return filename
```

────────────────────────── **code: python** ──────────────────────────

```python
1  def plot_x_section(ext='pdf', **kwargs):
2      fig, ax = plt.subplots(1, 1, figsize=(9, 3),
         ↪ constrained_layout=True)
3      # first subplot
4      title_text = "; ".join((f'{k}={v}' for k, v in
         ↪ kwargs.items()))
5      ax.set_title(f"X-Section ({title_text})")
6      modelmap = flopy.plot.PlotCrossSection(
7          model=gwf,
8          ax=ax,
9          line=kwargs,
10     )
11     k_values = npf.k.get_data()
12     pa = modelmap.plot_array(k_values, alpha=0.6)
13     quadmesh = modelmap.plot_bc("CHD")
14     linecollection = modelmap.plot_grid(lw=0.2, color="white")
15     minor_contours = modelmap.contour_array(
16         head_arr,
17         levels=np.arange(0, 25, .2),
18         linewidths=0.4,
19         colors='black'
20     )
21     contours = modelmap.contour_array(
22         head_arr,
23         head=head_arr,
24         levels=np.arange(0, 25, 1),
25         linewidths=0.8,
26         colors='black'
27     )
28     ax.clabel(contours, fmt="%.0f")
29     pv = modelmap.plot_vector(qx, qy, qz,
30                               headwidth=3, headlength=4,
                                  ↪ width=2e-3,
31                               pivot='mid', minshaft=2, hstep=4,
32                               scale=3,
```

```python
33                                      color='blue')
34      # plt.colorbar(pa, shrink=0.5, ax=ax)
35      filename = "_".join((f'{k}-{v}' for k, v in kwargs.items()))
36      saveas = f"./images/3_section_{filename}.{ext}"
37      plt.savefig(saveas)
38      plt.show()
39      return saveas
```
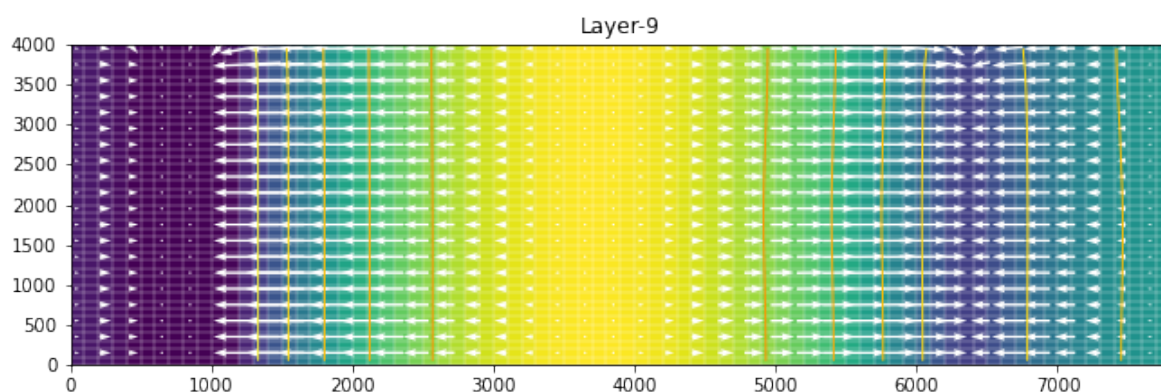
――――――――――――― code: python ―――――――――――――
```python
1  plot_plan(ext='png', layer=geolyr_subdivisions[0]-1)
```


The first Geo-Layer. Seems similar to the plot we got when we ran the model with well turned off, so we can say the effect isn't seen much on the top geolayer.

――――――――――――― code: python ―――――――――――――
```python
1  plot_plan(ext='png', layer=sum(geolyr_subdivisions[:2])-1)
```
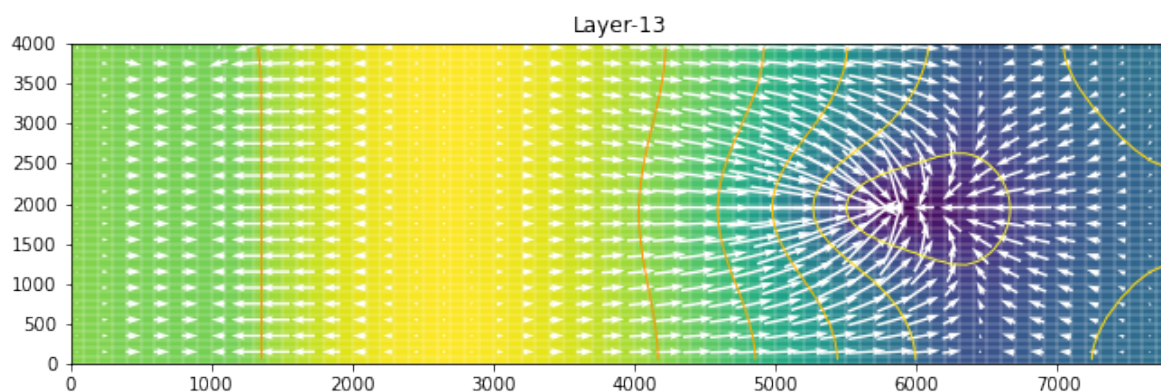

There seems to be some effect of well on the second geolayer.

――――――――――――― code: python ―――――――――――――
```python
1  lyr_index = sum(map(lambda b: b > well_bottom, bot))
2  plot_plan(ext='png', layer=lyr_index)
```
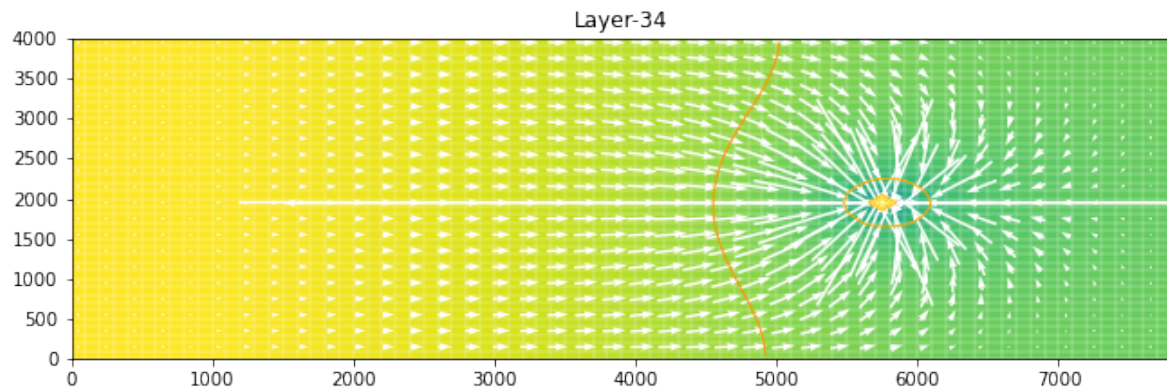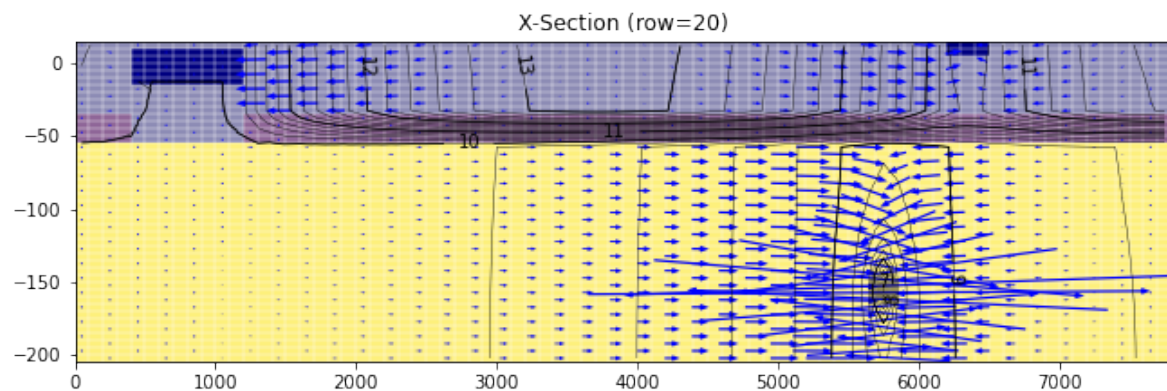
So the plot at the location of well bottom shows the cone of depression and the maximum effect.

```python
plot_x_section(ext='png', row=20)
```


This section shows the effect of well and how it is mostly in the 3rd geo layer, as well as how the effect extends to the river farther away but not the trout stream. We can see the isolines move towards the river through break in the clay layer showing the effect.

```python
plot_x_section(ext='png', column=60)
```
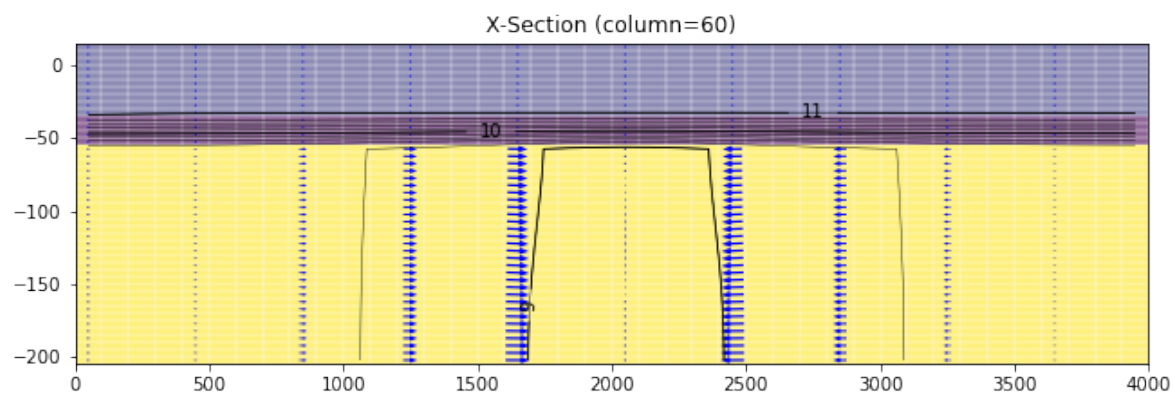

This column section doesn't show much except it's pretty much symmetric.

2.5 Zone Budget and effect of Well
Now let's define zone for trout stream, will use top 2 layers.

Here ZONE$_2$ will be the trout stream and everything else will be ZONE$_1$.

```python
zones = np.ones((NLay, NR, NC), dtype=int)
for p in get_grid_points(stream, layers=[0, 1]):
    zones[p] = 2
```

we can make a zonebudget model from the previous modflow model.

```python
bm = gwf.output.zonebudget(zones)

bm.change_model_name(name)
bm.change_model_ws(ws)
```

There seems to be a bug in write input method for budget model as I had to edit the output zbnam file to add grb file manually. So only run exisiting model files, don't write it unless you change the zone arr, in that case write it and manually add grb file again.

```python
bm.write_input()
bm.run_model(exe_name='modflow-zbud6')
```

NOTE: The `bm.write_input()` has a bug and doesn't write the grb file and it'll end up being an error. I brought it to the attention of the developers and they've fixed it. But it's not yet available to the pip version 3.3.5. If the next version isn't yet released when you're running this code, then you can install the git's latest version with `pip install --upgrade git+https://github.com/modflowpy/flopy.git` in your terminal. Or manually edit it.

The .zbnam file must have following contents:

```
BEGIN ZONEBUDGET
   ZON    3_water_wd.zon
   BUD    3_water_wd.bud
   GRB    3_water_wd.dis.grb
END ZONEBUDGET
```

If it doesn't have it, then either do latest install from github or edit the content of the file yourself and remove the `bm.write_input()` line from the code so it won't overwrite your edits. You have to write inputs if you change zones, or your grid settings and edit the file manually again. I'd recommend just installing the latest code from github so you don't have to.

With that we can get the water budget.

```python
bm.get_budget()
```

Without well: 29121.5759986 ft^3/day = 0.33705527776157407 cfs

With well: 20695.35801522 ft^3/day = 0.2395296066576389 cfs

Difference 0.337 - 0.240 = 0.0975 ft^3/sec

Hence the overall reduction is less than the 5% of the index flow.

Also if we also look at the reduction in the recharge to the river, we can sum both of them we get:
$5.329 \times 10^4 - 2.459 \times 10^4 + 2.912 \times 10^4 - 2.070 \times 10^4 = 37124.77086423 \, ft^3/day$

OR 192.86 GPM

Which is close to our well$_{rate}$ = -38499.8688 ft^3/day (-200 GPM).

3. Export to vtk format

We can also export the heads data we obtained from the simulation and then visualize it using external tools like Paraview.

```python
import os
from flopy.export import vtk
vtk.export_heads(sim.get_model(), os.path.join(ws, head_file), ws,
    smooth=False, kstpkper=[(0,0)], point_scalars=False,
    nanval=-1e30)
```