

1 Bluebonnet: Scaling solutions for production analysis 2 from unconventional oil and gas wells

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7 Summary

8 Unconventional oil and gas wells are only productive due to extensive hydraulic fracturing
9 treatments. Therefore, the character of their production over time is greatly influenced by
10 engineering decisions. However, it can be difficult to separate the engineering decisions from
11 the effects due to fluid properties. Also, during production these wells might be producing oil,
12 gas, and water simultaneously, with each phase interacting with the others. Numerical tools
13 are necessary to fully capture the effects of fluid properties on production.

14 Bluebonnet is a Python package that uses dimensionally scaled solutions of a pressure diffu-
15 sivity equation to analyze, history-match, and forecast production of tight-oil and shale gas
16 wells. Bluebonnet has been developed to help researchers and petroleum engineers analyzing
17 production data from unconventional (shale gas and tight oil) wells. It provides the user with
18 a set of tools to evaluate production performance of tight-oil and shale gas wells. These tools
19 provide the following functionality:

1. `bluebonnet.fluids`: pressure-volume-temperature properties for oil, water, and gas phases.
2. `bluebonnet.flow`: physics-based production curves and hydrocarbon recovery factors.
3. `bluebonnet.forecast`: fits and forecasts for unconventional production.

24 The `fluids` submodule estimates the formation volume factors, solubility ratios, and viscosity
25 for the oil, water and gas phases given the reservoir temperature, oil API gravity, gas specific
26 gravity, and initial gas/oil ratio.

27 The `flow` submodule solves the pressure diffusivity equation to provide estimates of the
28 hydrocarbon production over time and the hydrocarbon recovery factors. This module allows
29 the user to estimate production for shale gas wells using scaled solutions of the single-phase real
30 gas diffusivity equation (Male, 2015; Patzek et al., 2013). In addition, this module simulates
31 production for tight-oil and gas condensate wells using a two-phase scaled solution of the
32 pressure diffusivity equation (Ruiz Maraggi et al., 2022a). The `flow` submodule also allows
33 users to capture production variations due to changes in bottomhole pressure.

34 The `forecast` submodule performs history matches and forecasts the production of unconven-
35 tional wells using the scaling solutions present in the `flow` module. The `forecast` submodule
36 also allows users to history-match and forecast production of wells subject to variable bottom-
37 hole pressure conditions using a modification of the approach developed by Ruiz Maraggi et al.
38 (2022b).

39 Statement of need

40 Bluebonnet is a Python package using petroleum engineering methods to perform production
41 analysis of hydrofractured wells. Parts of this code were first developed to assist in determining
42 U.S. shale gas reserves (Male, 2019; Patzek et al., 2013).

43 There are no free open-source tools that use physics-based scaled flow solutions of the diffusivity
44 equation to perform decline-curve and rate-transient analysis for unconventional reservoirs
45 like bluebonnet. The goal for producing this software package is to provide researchers and
46 reservoir engineers with a free and open source tool suitable to analyze production from
47 unconventional (tight oil and shale gas) reservoirs.

48 The present library can be used for the following tasks:

- 49 1. Estimate fluid properties of reservoir fluids.
- 50 2. Build type curves and recovery factors for shale gas and tight-oil reservoirs.
- 51 3. History-match and forecast the production of shale gas and tight-oil wells.
- 52 4. Perform Rate-transient analysis (rate-time-pressure) of unconventional reservoirs.

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59 Deniz Ertas.

60 This project relies on the following open-source Python packages: NumPy (Harris et al., 2020;
61 Walt et al., 2011), SciPy (Virtanen et al., 2020), matplotlib (Hunter, 2007), and pandas
62 (McKinney, 2010).

63 The authors would like to thank the Society of Petroleum Engineers (SPE) for providing open
64 access to production data from unconventional wells through the SPE Data Repository, Data
65 Set 1 (Petroleum Engineers, 2021) used to illustrate the application of this package.

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