S3 Testing Iacono-Marziano et al 2012

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1 This notebook assesses the outputs of VESIcal for the Iacono-Marziano model.

This notebook uses the Excel spreadsheet entitled: "S3 Testing Iacono-Marziano et al 2012.xlsx" - Test 1 compares the experimental pressures for the H₂O-only experiments in the calibration dataset of Iacono-Marziano to the saturation pressures calculated in VESIcal using the "IaconoMarzianoWater" model. - Test 2 compares the experimental pressures for the H₂O-CO₂ experiments in the calibration dataset to the saturation pressures calculated in VESIcal for the "IaconoMarziano" model. - A lot of the scatter in the regression lines shown in Test 1 and Test 2 is experimental noise. For Test 3, major and volatile element concentrations and temperatures for this experimental dataset were entered into the saturation pressure web calculator hosted at http://calcul-isto.cnrs-orleans.fr/ (provided by Iacono-Marziano et al., 2012). These saturation pressures are compared to those from VESIcal for the "IaconoMarziano" model. - Test 4 compares saturation pressures obtained from the web calculator at http://calcul-isto.cnrsorleans.fr/ to VESIcal outputs for a synthetic dataset where composition, temperature, and volatile contents are varied - Test 5 compares dissolved volatiles calculated using the web app to those from VESIcal for a synthetic dataset with variable X_{H2O} .

```
[1]: import sys
    sys.path.insert(0, '../../../')

import VESIcal as v
    import matplotlib.pyplot as plt
    import numpy as np
    import pandas as pd
    from IPython.display import display, HTML
    import pandas as pd
    import matplotlib as mpl
    import seaborn as sns
    from sklearn.linear_model import LinearRegression
    from sklearn.metrics import r2_score
    import statsmodels.api as sm
    from statsmodels.sandbox.regression.predstd import wls_prediction_std
    //matplotlib inline
```

[2]:

```
sns.set(style="ticks", context="poster",rc={"grid.linewidth": 1,"xtick.major.
→width": 1,"ytick.major.width": 1, 'patch.edgecolor': 'black'})
plt.style.use("seaborn-colorblind")
plt.rcParams["font.size"] =12
plt.rcParams["mathtext.default"] = "regular"
plt.rcParams["mathtext.fontset"] = "dejavusans"
plt.rcParams['patch.linewidth'] = 1
plt.rcParams['axes.linewidth'] = 1
plt.rcParams["xtick.direction"] = "in"
plt.rcParams["ytick.direction"] = "in"
plt.rcParams["ytick.direction"] = "in"
plt.rcParams["xtick.major.size"] = 6 # Sets length of ticks
plt.rcParams["ytick.major.size"] = 4 # Sets length of ticks
plt.rcParams["ytick.labelsize"] = 12 # Sets size of numbers on tick marks
plt.rcParams["xtick.labelsize"] = 12 # Sets size of numbers on tick marks
plt.rcParams["axes.titlesize"] = 14 # Overall title
plt.rcParams["axes.labelsize"] = 14 # Axes labels
plt.rcParams["legend.fontsize"] = 14
```

2 Test 1 and 2 - Comparing experimental pressures to VESIcal saturation pressures for H_2O -only experiments and mixed H_2O - CO_2O experiments

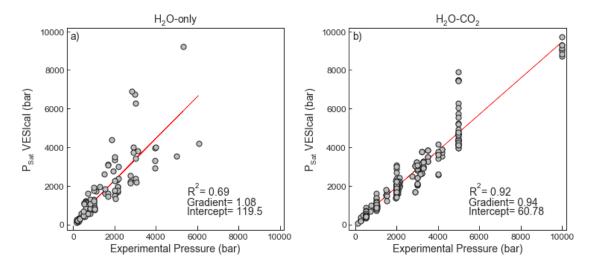
```
[3]: # This loads the calibration dataset of Iacono-Marziano et al. 2012 for
      →water-only experiments, and calculates saturation pressures based on the
      \rightarrowmajor element contents, temperature, and water content for H2O only
      \rightarrow experiments.
     myfile_H2Ocal = v.BatchFile('S3_Testing_Iacono-Marziano_et_al_2012.xlsx', _
      ⇔sheet_name='Calibration_H20')
     data_H2Ocal = myfile_H2Ocal.get_data()
     satPs_wtemps_Iacono_H2Ocal= myfile_H2Ocal.
     →calculate_saturation_pressure(temperature="Temp", __
      →model='IaconoMarzianoWater')
     # This loads the calibration dataset of Iacono-Marziano et al. 2012 for mixed
     → fluids and calculates saturation pressures
     myfile_cal = v.BatchFile('S3_Testing_Iacono-Marziano_et_al_2012.xlsx', u
     ⇔sheet_name='Calibration_H20C02')
     data_cal = myfile_cal.get_data()
     satPs_wtemps_Iacono_cal= myfile_cal.
      →calculate_saturation_pressure(temperature="Temp", model='IaconoMarziano')
```

/opt/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:2: RuntimeWarning: Total iron column FeOT detected. This column will be treated as FeO. If Fe2O3 data are not given, Fe2O3 will be 0.0. In future, an option to calcualte FeO/Fe203 based on fO2 will be implemented.

```
[4]: # This calculating a linear regression, and plots experimental pressures vs.
     \hookrightarrow saturation pressures (all data)
     X_Test1=satPs_wtemps_Iacono_H2Ocal['Press']
    Y_Test1=satPs_wtemps_Iacono_H2Ocal['SaturationP_bars_VESIcal']
    mask_Test1 = (X_Test1>-1) & (Y_Test1>-1) # This gets rid of Nans
    X_Test1noNan=X_Test1[mask_Test1].values.reshape(-1, 1)
    Y_Test1noNan=Y_Test1[mask_Test1].values.reshape(-1, 1)
    lr=LinearRegression()
    lr.fit(X Test1noNan, Y Test1noNan)
    Y_pred_Test1=lr.predict(X_Test1noNan)
    fig, (ax1, ax2) = plt.subplots(1,2, figsize=(12,5)) # adjust dimensions of
     → figure here
    ax1.plot(X_Test1noNan,Y_pred_Test1, color='red', linewidth=0.5, zorder=1) #__
     → This plots the best fit line
    ax1.scatter(satPs wtemps Iacono H2Ocal['Press'],
     ⇒satPs_wtemps_Iacono_H2Ocal['SaturationP_bars_VESIcal'], s=50, __
     →edgecolors='k', facecolors='silver', marker='o', zorder=5)
     # This bit plots the regression parameters on the graph
    I='Intercept= ' + str(np.round(lr.intercept_, 1))[1:-1]
    G='Gradient= ' + str(np.round(lr.coef_, 2))[2:-2]
    R='R$^2$= ' + str(np.round(r2_score(Y_Test1noNan, Y_pred_Test1), 2))
    ax1.text(5500, 1500, R, fontsize=14)
    ax1.text(5500, 1000, G, fontsize=14)
    ax1.text(5500, 500, I, fontsize=14)
    ################ Mixed H20 CO2 experiments
    X_Test2=satPs_wtemps_Iacono_cal['Press']
    Y Test2=satPs wtemps Iacono cal['SaturationP bars VESIcal']
    mask_Test2 = (X_Test2>-1) & (Y_Test2>-1) # This gets rid of Nans
    X_Test2noNan=X_Test2[mask_Test2].values.reshape(-1, 1)
    Y_Test2noNan=Y_Test2[mask_Test2].values.reshape(-1, 1)
    lr=LinearRegression()
    lr.fit(X_Test2noNan,Y_Test2noNan)
    Y_pred_Test2=lr.predict(X_Test2noNan)
    ax2.plot(X_Test2noNan,Y_pred_Test2, color='red', linewidth=0.5, zorder=1) #__
     \rightarrow This plots the best fit line
    ax2.scatter(satPs_wtemps_Iacono_cal['Press'],__
     ⇒satPs_wtemps_Iacono_cal['SaturationP_bars_VESIcal'], s=50, edgecolors='k',⊔

→facecolors='silver', marker='o', zorder=5)
     # This bit plots the regression parameters on the graph
```

```
I='Intercept= ' + str(np.round(lr.intercept_, 2))[1:-1]
G='Gradient= ' + str(np.round(lr.coef_, 2))[2:-2]
R='R$^2$= ' + str(np.round(r2_score(Y_Test2noNan, Y_pred_Test2), 2))
ax2.text(5500, 500, I, fontsize=14)
ax2.text(5500, 1000, G, fontsize=14)
ax2.text(5500, 1500, R, fontsize=14)
ax1.set_xlabel('Experimental Pressure (bar)', fontsize=14)
ax1.set ylabel('P$ {Sat}$ VESIcal (bar)', fontsize=14)
ax2.set_xlabel('Experimental Pressure (bar)', fontsize=14)
ax2.set ylabel('P$ {Sat}$ VESIcal (bar)', fontsize=14)
ax1.set_xticks([0, 2000, 4000, 6000, 8000, 10000])
ax1.set_yticks([0, 2000, 4000, 6000, 8000, 10000])
ax2.set_xticks([0, 2000, 4000, 6000, 8000, 10000])
ax2.set_yticks([0, 2000, 4000, 6000, 8000, 10000])
ax1.set_xlim([-300, 10200])
ax1.set_ylim([-300, 10200])
ax2.set_xlim([-300, 10200])
ax2.set_ylim([-300, 10200])
plt.subplots_adjust(left=0.125, bottom=None, right=0.9, top=None, wspace=0.3,_
→hspace=None)
ax1.text(-150, 9600, 'a)', fontsize=14)
ax2.text(-150, 9600, 'b)', fontsize=14)
ax1.set_title('H$_{2}$0-only', fontsize=14)
ax2.set_title('H$_{2}$0-CO$_2$', fontsize=14)
fig.savefig('IaconMarziano_Test1and2.png', transparent=True)
```



3 Test 3 - Comparing Saturation pressures from the web app to VESIcal for compositions in the calibration dataset

• The major and volatile element concentrations and temperatures from the calibration dataset were used to calculate saturation pressures in the web app, and compared to those in VESIcal

4 Test 4 - Comparing Saturation pressures from the web app to VESIcal for synthetic data

• A synthetic dataset varying major element compositions, temperature, and volatile contents was run through the web app and VESIcal

satP = self.models[1].calculate_saturation_pressure(sample=sample,**kwargs)

4.1 Plot for Test 3 and Test 4

```
fig, (ax1, ax2) = plt.subplots(1,2, figsize=(12,5)) # adjust dimensions of figure here

# Comparison of web app and VESIcal for calibration dataset

X_Test3=satPs_wtemps_Iacono_web['App calculator P bar'] # Convert MPa from their supplement to bars

Y_Test3=satPs_wtemps_Iacono_web['SaturationP_bars_VESIcal']

mask_Test3 = (X_Test3>-1) & (Y_Test3>-1) #& (XComb<7000) # This gets rid of Nans

X_Test3noNan=X_Test3[mask_Test3].values.reshape(-1, 1)

Y_Test3noNan=Y_Test3[mask_Test3].values.reshape(-1, 1)

lr=LinearRegression()

lr.fit(X_Test3noNan,Y_Test3noNan)

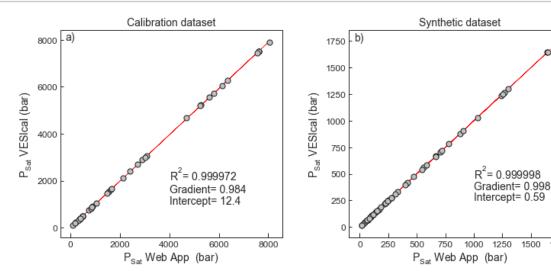
Y_pred_Test3=lr.predict(X_Test3noNan)
```

```
ax1.plot(X_Test3noNan,Y_pred_Test3, color='red', linewidth=0.5, zorder=1) #__
→ This plots the best fit line
ax1.scatter(satPs_wtemps_Iacono_web['App calculator P bar'],__
⇒satPs_wtemps_Iacono_web['SaturationP_bars_VESIcal'], s=50, edgecolors='k', __

¬facecolors='silver', marker='o', zorder=5)
# This bit plots the regression parameters on the graph
I='Intercept= ' + str(np.round(lr.intercept_, 2))[1:-1]
G='Gradient= ' + str(np.round(lr.coef_, 3))[2:-2]
R='R$^2$= ' + str(np.round(r2_score(Y_Test3noNan, Y_pred_Test3), 6))
ax1.text(4000, 1000, I, fontsize=14)
ax1.text(4000,1500, G, fontsize=14)
ax1.text(4000, 2000, R, fontsize=14)
# Comparison of web app and VESIcal for the synthetic dataset
X_Test4=satPs_wtemps_Iacono_synweb['Press']
Y_Test4=satPs_wtemps_Iacono_synweb['SaturationP_bars_VESIcal']
mask_Test4 = (X_Test4>-1) & (Y_Test4>-1) #8 (XComb<7000) # This gets rid of Nans
X_Test4noNan=X_Test4[mask_Test4].values.reshape(-1, 1)
Y_Test4noNan=Y_Test4[mask_Test4].values.reshape(-1, 1)
lr=LinearRegression()
lr.fit(X_Test4noNan,Y_Test4noNan)
Y_pred_Test4=lr.predict(X_Test4noNan)
ax2.plot(X_Test4noNan,Y_pred_Test4, color='red', linewidth=0.5, zorder=1) #__
→ This plots the best fit line
ax2.scatter(X_Test4noNan, Y_Test4noNan, s=50, edgecolors='k',_

→facecolors='silver', marker='o', zorder=5)
I='Intercept= ' + str(np.round(lr.intercept_, 2))[1:-1]
G='Gradient= ' + str(np.round(lr.coef_, 3))[2:-2]
R='R$^2$= ' + str(np.round(r2_score(Y_Test4noNan, Y_pred_Test4), 6))
ax2.text(1000, 250, I, fontsize=14)
ax2.text(1000, 350, G, fontsize=14)
ax2.text(1000, 450, R, fontsize=14)
ax1.set_xlabel('P$_{Sat}$ Web App (bar)', fontsize=14)
ax1.set_ylabel('P$_{Sat}$ VESIcal (bar)', fontsize=14)
ax2.set_xlabel('P$_{Sat}$ Web App (bar)', fontsize=14)
ax2.set_ylabel('P$_{Sat}$ VESIcal (bar)', fontsize=14)
ax1.set_yticks([0, 2000, 4000, 6000, 8000])
plt.subplots_adjust(left=0.125, bottom=None, right=0.9, top=None, wspace=0.3,__
→hspace=None)
ax1.set_title('Calibration dataset', fontsize=14)
ax2.set_title('Synthetic dataset', fontsize=14)
```

```
ax1.text(-200, 8000, 'a)', fontsize=14)
ax2.text(-50, 1750, 'b)', fontsize=14)
fig.savefig('IaconoMarziano_Test3and4.png', transparent=True)
```



Test 5 - Comparing dissolved volatiles from the web app to VESIcal

1500 1750

• The web app was used to calculate dissolved volatiles for a synthetic dataset with variable X_{H2O} contents

```
[8]: myfile synweb_cv = v.BatchFile('S3_Testing_Iacono-Marziano_et_al_2012.xlsx', __
      ⇒sheet name='Calculate Dissolved Volatiles') # This sheet has the dissolved
     →volatiles calculated using the web calculator
     data_synweb_cv = myfile_synweb_cv.get_data()
     dissolved syn = myfile synweb cv.
      →calculate_dissolved_volatiles(temperature="Temp", pressure="Press", __
      →X_fluid="XH20", print_status=True, model='IaconoMarziano')
```

```
[9]: fig, (ax1, ax2) = plt.subplots(1,2, figsize=(12,5)) # adjust dimensions of
      \rightarrow figure here
     # H20
     ###########
     X_syn1=dissolved_syn['H20(wt%)WebApp'].values.reshape(-1, 1)
     Y_syn1=dissolved_syn['H20_liq_VESIcal'].values.reshape(-1, 1)
     lr=LinearRegression()
     lr.fit(X_syn1,Y_syn1)
     Y_pred_syn1=lr.predict(X_syn1)
```

```
ax1.plot(X_syn1,Y_pred_syn1, color='red', linewidth=0.5, zorder=1) # This plots_
→ the best fit line
ax1.scatter(dissolved_syn['H2O(wt%)WebApp'], dissolved_syn['H2O_liq_VESIcal'], _
⇒s=50, edgecolors='k', facecolors='silver', marker='o', zorder=5)
# This bit plots the regression parameters on the graph
I='Intercept= ' + str(np.round(lr.intercept_, 3))[1:-1]
G='Gradient= ' + str(np.round(lr.coef_, 3))[2:-2]
R='R$^2$= ' + str(np.round(r2_score(Y_syn1, Y_pred_syn1), 5))
ax1.text(2, 0.5, I, fontsize=14)
ax1.text(2, 0.75, G, fontsize=14)
ax1.text(2, 1, R, fontsize=14)
# CO2
X_syn2=dissolved_syn['CO2(ppm)WebApp'].values.reshape(-1, 1)
Y_syn2=10000*dissolved_syn['CO2_liq_VESIcal'].values.reshape(-1, 1)
lr=LinearRegression()
lr.fit(X_syn2,Y_syn2)
Y_pred_syn2=lr.predict(X_syn2)
ax2.plot(X_syn2,Y_pred_syn2, color='red', linewidth=0.5, zorder=1) # This plots_
\rightarrowthe best fit line
ax2.scatter(dissolved_syn['CO2(ppm)WebApp'],_
→10000*dissolved_syn['CO2_liq_VESIcal'], s=50, edgecolors='k', __
# This bit plots the regression parameters on the graph
I='Intercept= ' + str(np.round(lr.intercept_, 2))[1:-1]
G='Gradient= ' + str(np.round(lr.coef_, 3))[2:-2]
R='R^2='+ str(np.round(r2_score(Y_syn2, Y_pred_syn2), 5))
plt.text(800, 200, I, fontsize=14)
plt.text(800, 300, G, fontsize=14)
plt.text(800, 400, R, fontsize=14)
#########################
ax1.set_xlabel('H$_2$0 Web App (wt%)', fontsize=14)
ax1.set_ylabel('H$_2$0 VESIcal (wt%)', fontsize=14)
ax2.set_xlabel('CO$_2$ Web App (ppm)', fontsize=14)
ax2.set_ylabel('CO$_2$ VESIcal (ppm)', fontsize=14)
plt.subplots_adjust(left=0.125, bottom=None, right=0.9, top=None, wspace=0.3,__
→hspace=None)
```

```
ax1.set_yticks([0, 1, 2, 3])
ax2.set_yticks([0, 400, 800, 1200, 1600])
ax2.set_xticks([0, 400, 800, 1200, 1600])
ax1.text(-0.2,3.45, 'a)', fontsize=14)
ax2.text(-70, 1520, 'b)', fontsize=14)
fig.savefig('IaconoMarziano_Test5.png', transparent=True)
```

