### HW 18-19

## **Imports**

```
import warnings
import numpy as np
import pandas as pd
from matplotlib import pyplot as plt
warnings.filterwarnings("ignore")
```

### **Bias-Variance Trade**

```
In [2]: # It's the same dataset for both HW 18 and HW 19
    data: np.ndarray = pd.read_csv("bias-variance-trade.txt", delimiter=" ", header=None).values.astype(np.floax1 = data[:, 0]
    y1 = data[:, 1]
    x2 = data[:, 2]
    y2 = data[:, 3]
```

### **HW 18**

In this problem, we will use the data in bias variance trade.txt to practice using cross-validation in model evaluation. The first column is one sample of the independent variable x, while the second is the dependent variable y that corresponds to those points. The third and fourth columns are the same for a different sample of the independent variable. You will want to complete the Bias-Variance Tradeoff problem before proceeding.

```
In [3]: # (a) Fit the data from the first two columns using Legendre polynomials of orders 1 through 10. # For each fit, preform leave—one—out cross validation and report the average error to the point
```

```
# left out. Note that this means for each order of polynomial, you will perform 20 fits. If you
# have not found a shortcut in your programming language of choice already, do so before
# proceeding.
# a design matrix is created as follows:
def design_mat(k: int, vec) -> np.ndarray:
    """ copied from HW 16 """
    A = np.stack([np.polynomial.legendre.Legendre.basis(i)(vec) for i in range(k)]).T
    return A
optimal fit k = -1
optimal fit mean mse = np.inf
optimal test k = -1
optimal test resid = np.inf
for k in range(1,12):
    A = design mat(k, x1)
    test resids = []
    fit mses = []
    for i in range(A.shape[0]):
        A tmp = np.concatenate([A[:i], A[i+1:]])
        y1 tmp = np.concatenate([y1[:i], y1[i+1:]])
        A i = A[i]
        y1 i = y1[i]
        theta_hat, sse_crossval, _, _ = np.linalg.lstsq(A_tmp, y1_tmp)
        mse_crossval = sse_crossval/A_tmp.shape[0]
        y1 i hat = A i@theta hat
        test resid = abs(y1 i hat-y1 i)
        test resids.append(test resid)
        fit mses.append(mse crossval)
        print(f"With K={k}, leaving out row {i}, our fit MSE is {mse crossval} and test residual is {test
    mean test resid = np.mean(test resids)
    mean fit mse = np.mean(fit mses)
    print("="*20)
    print(f"Across all leave-one-out runs for K={k}:\n- Our average test residual is {mean test resid} and
    if mean test resid < optimal test resid:</pre>
        optimal test resid = mean test resid
        optimal test k = k
    if mean fit mse < optimal fit mean mse:</pre>
```

```
optimal_fit_mean_mse = mean_fit_mse
    optimal_fit_k = k

print("="*40)
print(f"For this dataset and bucket of models:\n- The ideal fit by test residual is K={optimal_test_k}, wh:
```

With K=1, leaving out row 0, our fit MSE is [0.15328997] and test residual is 0.3683567385819108 With K=1, leaving out row 1, our fit MSE is [0.15396343] and test residual is 0.34959616727527376 With K=1, leaving out row 2, our fit MSE is [0.15694634] and test residual is 0.250118431909715 With K=1, leaving out row 3, our fit MSE is [0.15894036] and test residual is 0.15059502042414732 With K=1, leaving out row 4, our fit MSE is [0.15816128] and test residual is 0.1956029581896502 With K=1, leaving out row 5, our fit MSE is [0.15939403] and test residual is 0.11664257965398506 With K=1, leaving out row 6, our fit MSE is [0.14927922] and test residual is 0.46465214080537465 With K=1, leaving out row 7, our fit MSE is [0.15916202] and test residual is 0.13507648407008474 With K=1, leaving out row 8, our fit MSE is [0.15400719] and test residual is 0.3483422065965093 With K=1, leaving out row 9, our fit MSE is [0.14367974] and test residual is 0.5726178588970473 With K=1, leaving out row 10, our fit MSE is [0.11978988] and test residual is 0.8976015485981512 With K=1, leaving out row 11, our fit MSE is [0.15811015] and test residual is 0.19819977213052942 With K=1, leaving out row 12, our fit MSE is [0.15923434] and test residual is 0.12961216405596582 With K=1, leaving out row 13, our fit MSE is [0.16004365] and test residual is 0.02476212313970115 With K=1, leaving out row 14, our fit MSE is [0.12549402] and test residual is 0.8316282785624198 With K=1, leaving out row 15, our fit MSE is [0.15999217] and test residual is 0.04053004277839256 With K=1, leaving out row 16, our fit MSE is [0.15987557] and test residual is 0.06304477114805646 With K=1, leaving out row 17, our fit MSE is [0.12401573] and test residual is 0.8492181757187022 With K=1, leaving out row 18, our fit MSE is [0.15958055] and test residual is 0.09937296837860246 With K=1, leaving out row 19, our fit MSE is [0.16002717] and test residual is 0.03070354114558615

Across all leave-one-out runs for K=1:

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- Our average test residual is 0.3058136986029903 and our average fit mse is 0.15164933987639292 With K=2, leaving out row 0, our fit MSE is [0.15314715] and test residual is 0.37165575671635476 With K=2, leaving out row 1, our fit MSE is [0.15389458] and test residual is 0.3568495593647903 With K=2, leaving out row 2, our fit MSE is [0.15690382] and test residual is 0.25719603872289126 With K=2, leaving out row 3, our fit MSE is [0.15858366] and test residual is 0.1797795411188805 With K=2, leaving out row 4, our fit MSE is [0.15815998] and test residual is 0.19654575632515747 With K=2, leaving out row 5, our fit MSE is [0.15920936] and test residual is 0.13184502612171367 With K=2, leaving out row 6, our fit MSE is [0.1477693] and test residual is 0.5176594230070772 With K=2, leaving out row 7, our fit MSE is [0.15914496] and test residual is 0.1333936927518242 With K=2, leaving out row 8, our fit MSE is [0.15366184] and test residual is 0.3607762675880889 With K=2, leaving out row 9, our fit MSE is [0.14347844] and test residual is 0.5762470727062117 With K=2, leaving out row 10, our fit MSE is [0.11926507] and test residual is 0.9128319014688485 With K=2, leaving out row 11, our fit MSE is [0.15793943] and test residual is 0.22539276271597036 With K=2, leaving out row 12, our fit MSE is [0.15923205] and test residual is 0.12833700728330633 With K=2, leaving out row 13, our fit MSE is [0.16001578] and test residual is 0.01617260335167495 With K=2, leaving out row 14, our fit MSE is [0.1248838] and test residual is 0.8422135078363235 With K=2, leaving out row 15, our fit MSE is [0.15987391] and test residual is 0.05943008292196647 With K=2, leaving out row 16, our fit MSE is [0.15984364] and test residual is 0.060842378276132765 With K=2, leaving out row 17, our fit MSE is [0.12399179] and test residual is 0.8489922258885867 With K=2, leaving out row 18, our fit MSE is [0.15957434] and test residual is 0.0971155342399018

Across all leave-one-out runs for K=2:

- Our average test residual is 0.31591279857081 and our average fit mse is 0.15142546475927457 With K=3, leaving out row 0, our fit MSE is [0.15271669] and test residual is 0.3924648657252174 With K=3, leaving out row 1, our fit MSE is [0.15389432] and test residual is 0.3568010896249205 With K=3, leaving out row 2, our fit MSE is [0.15685525] and test residual is 0.2599700498070736 With K=3, leaving out row 3, our fit MSE is [0.15840919] and test residual is 0.20528594167642836 With K=3, leaving out row 4, our fit MSE is [0.15815986] and test residual is 0.19647941184997675 With K=3, leaving out row 5, our fit MSE is [0.15920852] and test residual is 0.1316657318807541 With K=3, leaving out row 6, our fit MSE is [0.14766069] and test residual is 0.5213071237435812 With K=3, leaving out row 7, our fit MSE is [0.15905258] and test residual is 0.14417421238482922 With K=3, leaving out row 8, our fit MSE is [0.15354798] and test residual is 0.365935988040416 With K=3, leaving out row 9, our fit MSE is [0.14246386] and test residual is 0.6140163226212358 With K=3, leaving out row 10, our fit MSE is [0.11747056] and test residual is 0.9509603946508585 With K=3, leaving out row 11, our fit MSE is [0.15725316] and test residual is 0.3065827265993309 With K=3, leaving out row 12, our fit MSE is [0.15923074] and test residual is 0.128155192536844 With K=3, leaving out row 13, our fit MSE is [0.16001382] and test residual is 0.01463949723628729 With K=3, leaving out row 14, our fit MSE is [0.12350586] and test residual is 0.8733106332297803 With K=3, leaving out row 15, our fit MSE is [0.15984808] and test residual is 0.07434243366888657 With K=3, leaving out row 16, our fit MSE is [0.15981216] and test residual is 0.06719213124326431 With K=3, leaving out row 17, our fit MSE is [0.1217816] and test residual is 0.9034980786155655 With K=3, leaving out row 18, our fit MSE is [0.15957032] and test residual is 0.09706255024094389 With K=3, leaving out row 19, our fit MSE is [0.15992359] and test residual is 0.04750132003254135 \_\_\_\_\_

Across all leave-one-out runs for K=3:

- Our average test residual is 0.3325672847704368 and our average fit mse is 0.1510189412603708 With K=4, leaving out row 0, our fit MSE is [0.15265973] and test residual is 0.39679648677430895 With K=4, leaving out row 1, our fit MSE is [0.1535268] and test residual is 0.37465298715471207 With K=4, leaving out row 2, our fit MSE is [0.15673981] and test residual is 0.2666754937112475 With K=4, leaving out row 3, our fit MSE is [0.15821452] and test residual is 0.22433482295550272 With K=4, leaving out row 4, our fit MSE is [0.15800817] and test residual is 0.2083457893296134 With K=4, leaving out row 5, our fit MSE is [0.15918304] and test residual is 0.13843102161611387 With K=4, leaving out row 6, our fit MSE is [0.14733438] and test residual is 0.5374809471327835 With K=4, leaving out row 7, our fit MSE is [0.15897029] and test residual is 0.15213242214600303 With K=4, leaving out row 8, our fit MSE is [0.1533422] and test residual is 0.38014787367195696 With K=4, leaving out row 9, our fit MSE is [0.14229904] and test residual is 0.6211870500112732 With K=4, leaving out row 10, our fit MSE is [0.11372905] and test residual is 1.0301576672553059 With K=4, leaving out row 11, our fit MSE is [0.15533826] and test residual is 0.5389434948189504 With K=4, leaving out row 12, our fit MSE is [0.15914827] and test residual is 0.13682711015366988 With K=4, leaving out row 13, our fit MSE is [0.16000419] and test residual is 0.014593863528355722 With K=4, leaving out row 14, our fit MSE is [0.1225766] and test residual is 0.89783738293258

With K=4, leaving out row 15, our fit MSE is [0.15967182] and test residual is 0.1277300784548458 With K=4, leaving out row 16, our fit MSE is [0.15977849] and test residual is 0.07207802472483057 With K=4, leaving out row 17, our fit MSE is [0.12171614] and test residual is 0.9044496404799797 With K=4, leaving out row 18, our fit MSE is [0.15951591] and test residual is 0.10371617967644925 With K=4, leaving out row 19, our fit MSE is [0.15991922] and test residual is 0.046462230554837836

Across all leave-one-out runs for K=4:

- Our average test residual is 0.358649028354166 and our average fit mse is 0.15058379726283694 With K=5, leaving out row 0, our fit MSE is [0.14987762] and test residual is 0.34974521400947783 With K=5, leaving out row 1, our fit MSE is [0.15059597] and test residual is 0.3275373707391907 With K=5, leaving out row 2, our fit MSE is [0.15369363] and test residual is 0.198381092911881 With K=5, leaving out row 3, our fit MSE is [0.15355707] and test residual is 0.22620929848976423 With K=5, leaving out row 4, our fit MSE is [0.15420492] and test residual is 0.16197084229681497 With K=5, leaving out row 5, our fit MSE is [0.15525356] and test residual is 0.05722872345857338 With K=5, leaving out row 6, our fit MSE is [0.14624161] and test residual is 0.4829464578668591 With K=5, leaving out row 7, our fit MSE is [0.15503396] and test residual is 0.09028128438096136 With K=5, leaving out row 8, our fit MSE is [0.14801982] and test residual is 0.40015444004250644 With K=5, leaving out row 9, our fit MSE is [0.1309935] and test residual is 0.7549526557333004 With K=5, leaving out row 10, our fit MSE is [0.11041964] and test residual is 1.016612167384852 With K=5, leaving out row 11, our fit MSE is [0.15496346] and test residual is 0.3005085601796941 With K=5, leaving out row 12, our fit MSE is [0.15505107] and test residual is 0.08633875284622902 With K=5, leaving out row 13, our fit MSE is [0.15499648] and test residual is 0.09919490678442977 With K=5, leaving out row 14, our fit MSE is [0.12023683] and test residual is 0.8746169244938655 With K=5, leaving out row 15, our fit MSE is [0.15504153] and test residual is 0.14880707495360146 With K=5, leaving out row 16, our fit MSE is [0.15538508] and test residual is 0.0041145154114381015 With K=5, leaving out row 17, our fit MSE is [0.10596806] and test residual is 1.0669963995269627 With K=5, leaving out row 18, our fit MSE is [0.15530632] and test residual is 0.04232037607709728 With K=5, leaving out row 19, our fit MSE is [0.15524957] and test residual is 0.05878109688206885 \_\_\_\_\_

Across all leave-one-out runs for K=5:

- Our average test residual is 0.33738490772347846 and our average fit mse is 0.14600448513136435 With K=6, leaving out row 0, our fit MSE is [0.14889758] and test residual is 0.4043790402527211 With K=6, leaving out row 1, our fit MSE is [0.15052462] and test residual is 0.3303417257967663 With K=6, leaving out row 2, our fit MSE is [0.15330456] and test residual is 0.2351125622433497 With K=6, leaving out row 3, our fit MSE is [0.15341652] and test residual is 0.2491911268341379 With K=6, leaving out row 4, our fit MSE is [0.15418999] and test residual is 0.1621198664902348 With K=6, leaving out row 5, our fit MSE is [0.15522406] and test residual is 0.06146824300309153 With K=6, leaving out row 6, our fit MSE is [0.1461645] and test residual is 0.4886946106417849 With K=6, leaving out row 7, our fit MSE is [0.15503362] and test residual is 0.499092529526161583 With K=6, leaving out row 8, our fit MSE is [0.14763708] and test residual is 0.41707366706905663 With K=6, leaving out row 9, our fit MSE is [0.1307445] and test residual is 0.7609042663929086 With K=6, leaving out row 10, our fit MSE is [0.10432668] and test residual is 1.1611420197943936

With K=6, leaving out row 11, our fit MSE is [0.15444668] and test residual is 0.7682314919959743 With K=6, leaving out row 12, our fit MSE is [0.1550345] and test residual is 0.08684982766580746 With K=6, leaving out row 13, our fit MSE is [0.15487423] and test residual is 0.11959103519986855 With K=6, leaving out row 14, our fit MSE is [0.11506155] and test residual is 0.9965701686089126 With K=6, leaving out row 15, our fit MSE is [0.1549578] and test residual is 0.20256007930196804 With K=6, leaving out row 16, our fit MSE is [0.15537296] and test residual is 0.0005750691644094211 With K=6, leaving out row 17, our fit MSE is [0.10372332] and test residual is 1.1192240075626658 With K=6, leaving out row 18, our fit MSE is [0.15528487] and test residual is 0.04484659463325581 With K=6, leaving out row 19, our fit MSE is [0.15520155] and test residual is 0.06838827822717974

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#### Across all leave-one-out runs for K=6:

- Our average test residual is 0.3884094488070051 and our average fit mse is 0.145171059195411 With K=7, leaving out row 0, our fit MSE is [0.13830655] and test residual is 0.44911729716325 With K=7, leaving out row 1, our fit MSE is [0.13981516] and test residual is 0.3822344811516015 With K=7, leaving out row 2, our fit MSE is [0.14558356] and test residual is 0.13507336535529774 With K=7, leaving out row 3, our fit MSE is [0.12778471] and test residual is 0.9905640292024365 With K=7, leaving out row 4, our fit MSE is [0.14387103] and test residual is 0.23155605425266143 With K=7, leaving out row 5, our fit MSE is [0.14197843] and test residual is 0.3730932435412918 With K=7, leaving out row 6, our fit MSE is [0.13473289] and test residual is 0.5495709576816451 With K=7, leaving out row 7, our fit MSE is [0.14557425] and test residual is 0.12720497375877945 With K=7, leaving out row 8, our fit MSE is [0.14127077] and test residual is 0.3408132412587204 With K=7, leaving out row 9, our fit MSE is [0.12690693] and test residual is 0.6917170781084172 With K=7, leaving out row 10, our fit MSE is [0.10324949] and test residual is 1.111216695165346 With K=7, leaving out row 11, our fit MSE is [0.13923015] and test residual is 3.688801662500599 With K=7, leaving out row 12, our fit MSE is [0.14524915] and test residual is 0.14952608475109172 With K=7, leaving out row 13, our fit MSE is [0.14051139] and test residual is 0.45798851839494664 With K=7, leaving out row 14, our fit MSE is [0.10834296] and test residual is 0.9684952781098219 With K=7, leaving out row 15, our fit MSE is [0.12775394] and test residual is 2.0135875652723083 With K=7, leaving out row 16, our fit MSE is [0.1461467] and test residual is 0.04638176523166292 With K=7, leaving out row 17, our fit MSE is [0.10243019] and test residual is 1.069921767117001 With K=7, leaving out row 18, our fit MSE is [0.1459684] and test residual is 0.07816705819579488 With K=7, leaving out row 19, our fit MSE is [0.14525765] and test residual is 0.16611697674698933

### Across all leave-one-out runs for K=7:

- Our average test residual is 0.701057404647983 and our average fit mse is 0.13449821394002276 With K=8, leaving out row 0, our fit MSE is [0.13821815] and test residual is 0.43233357126706173 With K=8, leaving out row 1, our fit MSE is [0.13907587] and test residual is 0.36313802769340453 With K=8, leaving out row 2, our fit MSE is [0.14429316] and test residual is 0.1134994692213854 With K=8, leaving out row 3, our fit MSE is [0.11606616] and test residual is 1.9789049367033682 With K=8, leaving out row 4, our fit MSE is [0.14277774] and test residual is 0.21284641218432523 With K=8, leaving out row 5, our fit MSE is [0.13503842] and test residual is 0.6527402284890258

With K=8, leaving out row 7, our fit MSE is [0.14347513] and test residual is 0.18152534933246883 With K=8, leaving out row 8, our fit MSE is [0.13998702] and test residual is 0.3339990997251288 With K=8, leaving out row 9, our fit MSE is [0.12658837] and test residual is 0.6781970322309566 With K=8, leaving out row 10, our fit MSE is [0.09937845] and test residual is 1.14885825379254 With K=8, leaving out row 11, our fit MSE is [0.08841988] and test residual is 21.177700219241977 With K=8, leaving out row 12, our fit MSE is [0.14403424] and test residual is 0.1282104999721821 With K=8, leaving out row 13, our fit MSE is [0.13161338] and test residual is 0.824171137754102 With K=8, leaving out row 14, our fit MSE is [0.10802911] and test residual is 0.9947396570407225 With K=8, leaving out row 15, our fit MSE is [0.10676893] and test residual is 5.449274782703361 With K=8, leaving out row 16, our fit MSE is [0.14439757] and test residual is 0.09496260595507874 With K=8, leaving out row 17, our fit MSE is [0.14464051] and test residual is 0.05008205011599272 With K=8, leaving out row 19, our fit MSE is [0.14362153] and test residual is 0.17859480214951756

Across all leave-one-out runs for K=8:

- Our average test residual is 1.8398965888096452 and our average fit mse is 0.12810877315143115 With K=9, leaving out row 0, our fit MSE is [0.04628131] and test residual is 0.41687543097469454 With K=9, leaving out row 1, our fit MSE is [0.04811246] and test residual is 0.3141204821977115 With K=9, leaving out row 2, our fit MSE is [0.031703] and test residual is 0.8747127399518615 With K=9, leaving out row 3, our fit MSE is [0.03249302] and test residual is 2.8877583788779417 With K=9, leaving out row 4, our fit MSE is [0.04826212] and test residual is 0.3073320473494035 With K=9, leaving out row 5, our fit MSE is [0.04659367] and test residual is 0.5053442490495397 With K=9, leaving out row 6, our fit MSE is [0.05232239] and test residual is 0.0309741847041905 With K=9, leaving out row 7, our fit MSE is [0.0517591] and test residual is 0.12393213972131961 With K=9, leaving out row 8, our fit MSE is [0.05146676] and test residual is 0.15193802362637965 With K=9, leaving out row 9, our fit MSE is [0.04924574] and test residual is 0.2909947671189159 With K=9, leaving out row 10, our fit MSE is [0.04849241] and test residual is 0.39425693673607853 With K=9, leaving out row 11, our fit MSE is [0.05107347] and test residual is 5.944730477291114 With K=9, leaving out row 12, our fit MSE is [0.05095094] and test residual is 0.17980319619658006 With K=9, leaving out row 13, our fit MSE is [0.04716841] and test residual is 0.6308562049694412 With K=9, leaving out row 14, our fit MSE is [0.04020609] and test residual is 0.5964119612420273 With K=9, leaving out row 15, our fit MSE is [0.0340562] and test residual is 7.528773607093412 With K=9, leaving out row 16, our fit MSE is [0.05197828] and test residual is 0.09808558977341278 With K=9, leaving out row 17, our fit MSE is [0.03979902] and test residual is 0.6326289028255502 With K=9, leaving out row 18, our fit MSE is [0.05188273] and test residual is 0.10558859054746804 With K=9, leaving out row 19, our fit MSE is [0.04639177] and test residual is 0.4151026243960264 \_\_\_\_\_

Across all leave-one-out runs for K=9:

- Our average test residual is 1.1215110267321535 and our average fit mse is 0.046011943282771285 With K=10, leaving out row 0, our fit MSE is [0.04476118] and test residual is 0.34825491666457553 With K=10, leaving out row 1, our fit MSE is [0.04501707] and test residual is 0.2893649597798163 With K=10, leaving out row 2, our fit MSE is [0.02922672] and test residual is 1.1562704970135067

With K=10, leaving out row 3, our fit MSE is [0.01515446] and test residual is 7.881457285320115 With K=10, leaving out row 4, our fit MSE is [0.04568428] and test residual is 0.26302296541168124 With K=10, leaving out row 5, our fit MSE is [0.03585503] and test residual is 0.8306874280086476 With K=10, leaving out row 6, our fit MSE is [0.04808842] and test residual is 0.14145778661190728 With K=10, leaving out row 7, our fit MSE is [0.04720543] and test residual is 0.1928539785559728 With K=10, leaving out row 8, our fit MSE is [0.04840519] and test residual is 0.0709931569470788 With K=10, leaving out row 9, our fit MSE is [0.04670006] and test residual is 0.23220448164485452 With K=10, leaving out row 10, our fit MSE is [0.04682442] and test residual is 0.2877874084129701 With K=10, leaving out row 11, our fit MSE is [0.01684762] and test residual is 59.351307988029234 With K=10, leaving out row 12, our fit MSE is [0.04778627] and test residual is 0.13734852696269212 With K=10, leaving out row 13, our fit MSE is [0.0263886] and test residual is 1.731660610828235 With K=10, leaving out row 14, our fit MSE is [0.03759717] and test residual is 0.5698622397415452 With K=10, leaving out row 15, our fit MSE is [0.0130878] and test residual is 24.811585954322258 With K=10, leaving out row 16, our fit MSE is [0.04749468] and test residual is 0.17148618993884154 With K=10, leaving out row 17, our fit MSE is [0.03156236] and test residual is 0.758636229933148 With K=10, leaving out row 18, our fit MSE is [0.0480719] and test residual is 0.11037911546598976 With K=10, leaving out row 19, our fit MSE is [0.03834877] and test residual is 0.5721004697658572 

Across all leave-one-out runs for K=10:

- Our average test residual is 4.9954361094679465 and our average fit mse is 0.03800537095759345 With K=11, leaving out row 0, our fit MSE is [0.00701223] and test residual is 0.12408541397240291 With K=11, leaving out row 1, our fit MSE is [0.00677646] and test residual is 0.13013232220222026 With K=11, leaving out row 2, our fit MSE is [0.00713685] and test residual is 0.23080653389292416 With K=11, leaving out row 3, our fit MSE is [0.00747783] and test residual is 0.037883285156537605 With K=11, leaving out row 4, our fit MSE is [0.00711586] and test residual is 0.0943788184719673 With K=11, leaving out row 5, our fit MSE is [0.00500093] and test residual is 0.5324005555840042 With K=11, leaving out row 6, our fit MSE is [0.00644299] and test residual is 0.21195391677325875 With K=11, leaving out row 7, our fit MSE is [0.00730476] and test residual is 0.06885588374201584 With K=11, leaving out row 8, our fit MSE is [0.00685601] and test residual is 0.1327536730216144 With K=11, leaving out row 9, our fit MSE is [0.00735394] and test residual is 0.06183171865266912 With K=11, leaving out row 10, our fit MSE is [0.0052528] and test residual is 0.356320525271532 With K=11, leaving out row 11, our fit MSE is [0.0072613] and test residual is 11.762281911217348 With K=11, leaving out row 12, our fit MSE is [0.00742399] and test residual is 0.036296907701273196 With K=11, leaving out row 13, our fit MSE is [0.00747064] and test residual is 0.04574554668871414 With K=11, leaving out row 14, our fit MSE is [0.00675652] and test residual is 0.15897864757348912 With K=11, leaving out row 15, our fit MSE is [0.00746446] and test residual is 1.252130681609544 With K=11, leaving out row 16, our fit MSE is [0.00716579] and test residual is 0.09208359282144338 With K=11, leaving out row 17, our fit MSE is [0.00464293] and test residual is 0.3383257613512497 With K=11, leaving out row 18, our fit MSE is [0.00489747] and test residual is 0.250210328542111 With K=11, leaving out row 19, our fit MSE is [0.00724903] and test residual is 0.09485697961848333

\_\_\_\_\_

Across all leave-one-out runs for K=11:

- Our average test residual is 0.8006156501932402 and our average fit mse is 0.006703138983798042

For this dataset and bucket of models:

- The ideal fit by test residual is K=1, which has a mean test residual of 0.3058136986029903
- The ideal fit by fit mse is 11, which has a mean fit mse of 0.006703138983798042

```
In [4]: # (b) Identify the order of polynomial that yields the lowest average error. Is it the same as the # best fitting models from the previous problem? What does this tell you about predictive # modeling generally?

# As shown in the output from the cell above, the order of polynomial that yields the lowest # mse in modeling is the tenth-order polynomial (11 terms, k=11). However, the order of polynomial # that yields the lowest average test residual is the zero-th-order polynomial (1 term, k=1). # This highlights that while the Legendre polynomials were effective at fitting the training data # effectively, they did not effectively capture the underlying physics.
```

### HW 19

In this problem, we will revisit the data in bias variance trade.txt to demonstrate generalized aliasing. Remember that the first and third columns are samples of the independent variable x, while the second and fourth columns are samples of the dependent variable y that correspond to those points.

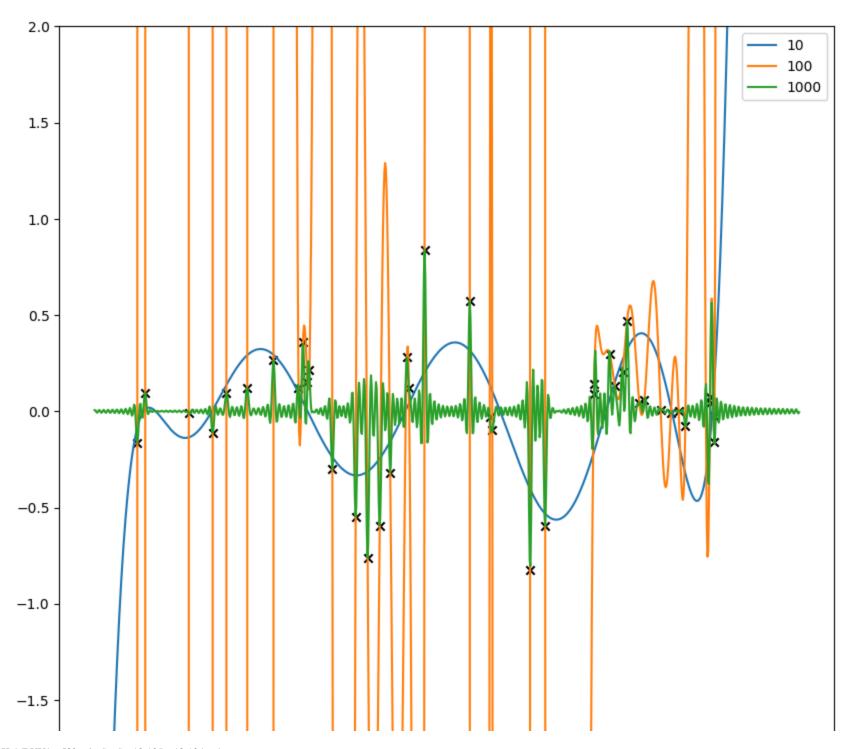
```
In [5]: # (a) Combine the two samples into one dataset and generate a system matrix assuming a basis of
        # sines and cosines with frequency that increases with column number (i.e. cos(x), cos(2x), cos(3x),
        # etc.). Your system matrix should have 1,000 columns. Do not print this matrix.
        # a design matrix is created as follows:
        def design mat(k: int, vec) -> np.ndarray:
            """ adapted from HW 18 """
            columns = []
            for i in range(k):
                con plus = i+2
                con_pmod = con_plus%2
                con split = con plus-con pmod
                constant = con split/2
                if i%2==0:
                     columns.append(np.sin(constant*vec))
                else:
                    columns.append(np.cos(constant*vec))
```

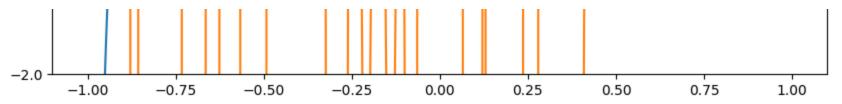
```
A = np.stack(columns).T
            return A
        x full = np.concatenate((x1, x2))
        y full = np.concatenate((y1, y2))
        A full = design mat(1000, x full)
        # we won't print the full matrix, but just the first 5 columns of the first five rows to get a picture of
        pd.DataFrame(A full[:10, :10])
Out[5]:
                  0
                                      2
                                               3
                                                                  5
                                                                                      7
                                                                                                 8
                                                                                                           9
                            1
        0 -0.160021 0.987114 -0.315917 0.948787 -0.463672 0.886007 -0.599476
                                                                                0.800393 -0.719830
                                                                                                     0.694150
        1 -0.397202  0.917731  -0.729049  0.684462  -0.940940
                                                            0.338573 -0.998012 -0.063024 -0.890874
                                                                                                    -0.454251
        2 -0.472548  0.881305  -0.832918  0.553396
                                                                     -0.921867 -0.387506 -0.629330
                                                  -0.995561
                                                            0.094116
                                                                                                    -0.777138
        3 -0.616901 0.787041 -0.971053 0.238866
                                                  -0.911614 -0.411047 -0.463902 -0.885886
                                                                                          0.181395 -0.983410
        4 -0.380407 0.924819
                             -0.703615
                                        0.710581
                                                  -0.921027 0.389499
                                                                     -0.999951
                                                                                0.009851 -0.928522
                                                                                                    -0.371278
           0.406080 0.913838
                               0.742182
                                        0.670199
                                                  0.950388
                                                            0.311068
                                                                      0.994818
                                                                                -0.101667
                                                                                          0.867818 -0.496883
            0.489518 0.871993
                               0.853712 0.520745
                                                  0.999345
                                                            0.036178
                                                                      0.889132 -0.457650
                                                                                          0.551290
                                                                                                   -0.834314
            0.128753 0.991677
                               0.255363 0.966845
                                                   0.377722
                                                            0.925919
                                                                                                     0.798764
        7
                                                                      0.493793
                                                                                0.869579
                                                                                          0.601645
           -0.827698
                                                            0.561173 -0.963556
                                                                                0.267507
                                                                                         -0.998532
                                                                                                    -0.054166
            0.065082 0.997880
                               0.129887 0.991529
                                                                                                    0.947447
                                                   0.194142 0.980973
                                                                      0.257574
                                                                                0.966259
                                                                                          0.319914
In [6]: # (b) Use your programming language of choice to create an integer list of terms that are roughly
        # equidistant in logarithmic space between 1 and 1,000. For each number of terms, produce
        # a fit to the combined data set. Plot a representative subsample of these fits with the data
        # used to produce them.
        integer list = np.logspace(1, 3, 3)
        x_for_plotting = np.linspace(-1, 1, 1000)
        plt.figure(figsize=(10, 10))
        plt.scatter(x full, y full, marker="x", color="black")
        for x int in integer list:
```

```
A_tmp = design_mat(int(x_int), x_full)
theta_hat, sse, rank, s = np.linalg.lstsq(A_tmp, y_full)
A_tmp = design_mat(int(x_int), x_for_plotting)
y_for_plotting = A_tmp@theta_hat
plt.plot(x_for_plotting, y_for_plotting, label=str(int(x_int)))

plt.legend()
plt.ylim(-2, 2)
```

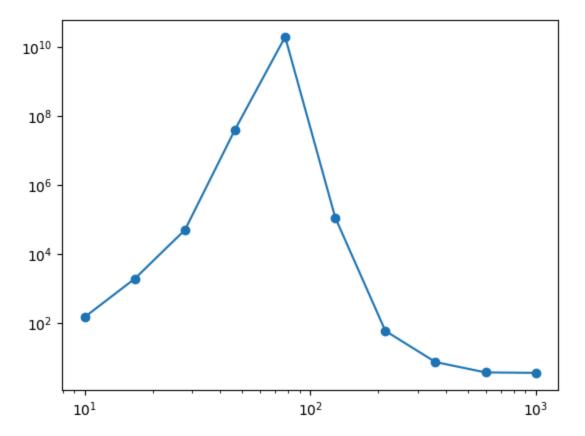
Out[6]: (-2.0, 2.0)





```
In [7]: # (c) Assuming that the true signal is characterized by f(x) = e^{-6x^2} \cos(4\pi x), calculate the Eu-
        # clidean norm of the difference between the true signal and each of your models for 100 equally
        # spaced samples between -1 and 1. Report the cost versus number of terms in a log-log plot.
        integer_list = np.logspace(1, 3, 10) # in this plot, we'll use more models
        x \text{ spaced} = \text{np.linspace}(-1, 1, 100)
        true y = np.exp(-6*(x spaced**2))*np.cos(4*np.pi*x spaced)
         costs = []
        for x int in integer list:
            x int = int(x int)
            A tmp = design mat(x int, x full)
            theta_hat, sse, rank, s = np.linalg.lstsq(A_tmp, y_full)
            A tmp = design mat(x int, x spaced)
            y_hat = A_tmp@theta_hat
            norm tmp = np.linalq.norm(np.abs(y hat-true y), ord=2)
             costs.append(norm tmp)
        plt.loglog(integer list, costs, marker='o') # plotting call from chatGPT
```

Out[7]: [<matplotlib.lines.Line2D at 0x112973b10>]



In [8]: # (d) Using the language of the General Aliasing Decomposition, explain the shape of the plot
# you produced in the prior part.

# In our plot we can observe double descent, with an interpolation threshold at 10^2.
# As we add additional/redundant parameters, they "absorb" the error to smooth out the pattern and
# better fit the underlying true relationship

# Acknowledgment

Work in this repository and with associated assignments and projects may be adapted or copied from similar files used in my prior academic and industry work (e.g., using a LaTeX file or Dockerfile as a starting point). Those files and any other work in this repository may have been developed with the help of LLM's like ChatGPT. For example, to provide context, answer questions, refine writing, understand function call syntax, and assist with repetitive tasks. In these cases, deliverables and

associated work reflect my best efforts to optimize my learning and demonstrate my capacity, while using available resources and LLM's to facilitate the process.

**ChatGPT Conversation**