

APPLYING COMPUTER VISION TO JET FLAVOR IDENTIFICATION

JOSH COGAN

MICHAEL KAGAN

EMANUEL STRAUSS

ARIEL SCHWARTZMAN

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NATIONAL
ACCELERATOR
LABORATORY

MOTIVATION

- Computer vision: high-dimensional data → usable information
 - Gender recognition, optical character recognition
 - Similar to a myriad of HEP problems—object identification
- Here, focus on calorimeter jet flavor classification
- Use linear methods on *all available* information to
 - Perform feature extraction/dimensional reduction
 - Use transparency of these methods to inspect them physically

SAMPLES STUDIED

- Event Generation: pp @ $\sqrt{s} = 8 \text{ TeV}$
 - W vs Light: Pythia $W(\mu\nu)W(qq)$ vs $W(\mu\nu)j$
 - Cross Checks Performed:
 - Pile up
 - Fast detector simulation (Loch)
 - Compare generators
- Calorimeter: $0.1 \times 0.1 \Delta\eta \times \Delta\phi$, $|\eta| < 2.5$
- C/A R=1.2 Jet Finding:
 - Save highest pT jet / event, $|\eta_j| < 2.5$
 - Trimmed (5% kt0.3)
 - Only keep towers that survived trimming
 - Study jets in 6 bins of subjet- ΔR
 - Jet Mass $> 55 \text{ GeV}$

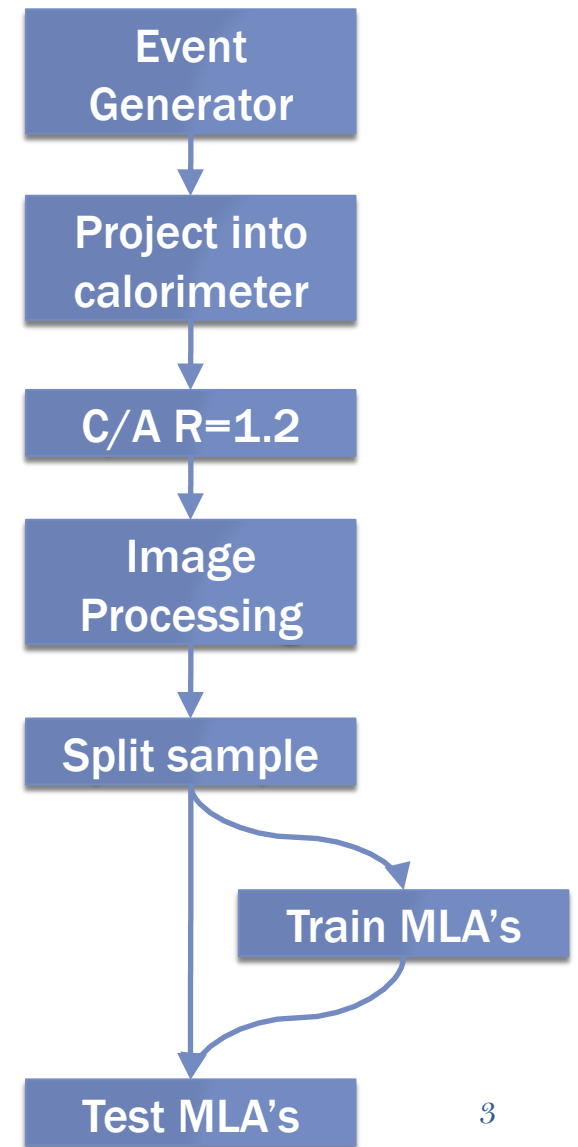
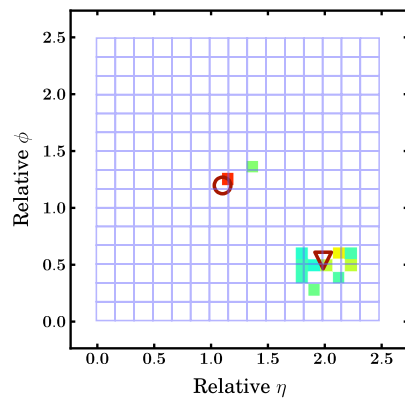
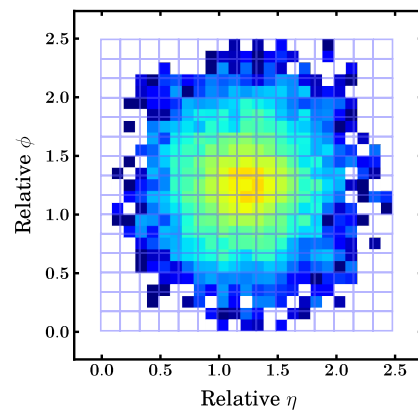


IMAGE PROCESSING



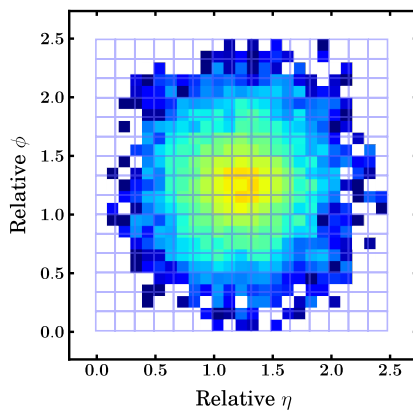
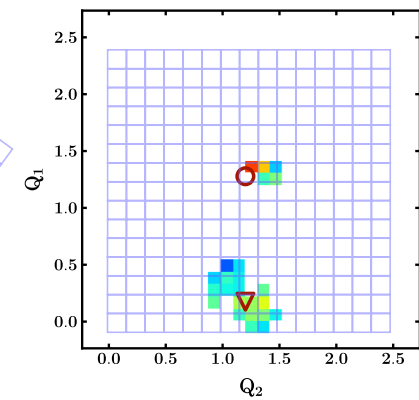
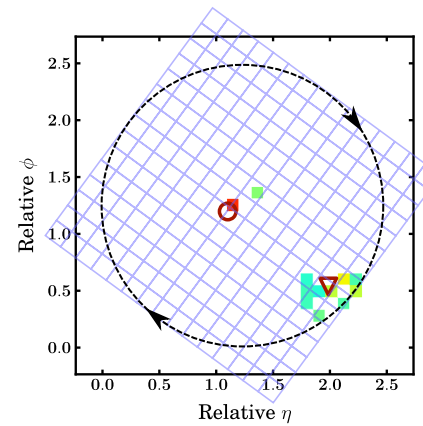
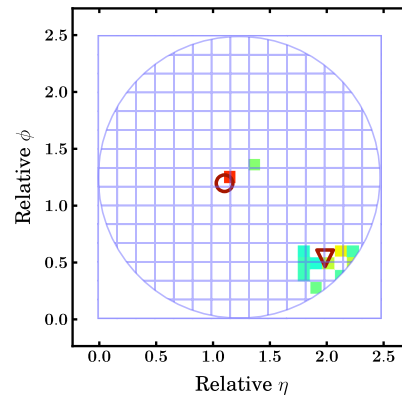
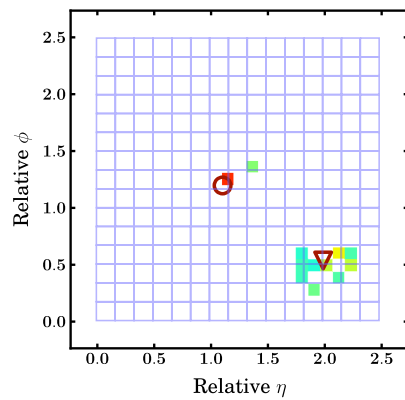
Example W jet



**Average of W jets
Not much info!**

IMAGE PROCESSING

- Center and rotate jet-images before training MVA
- Introduces small smearing, but huge gain in discrimination!

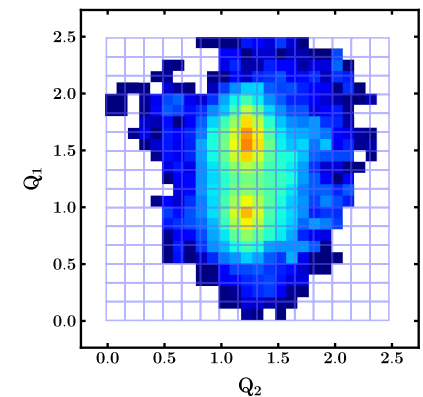


**Average of
unrotated W jet**

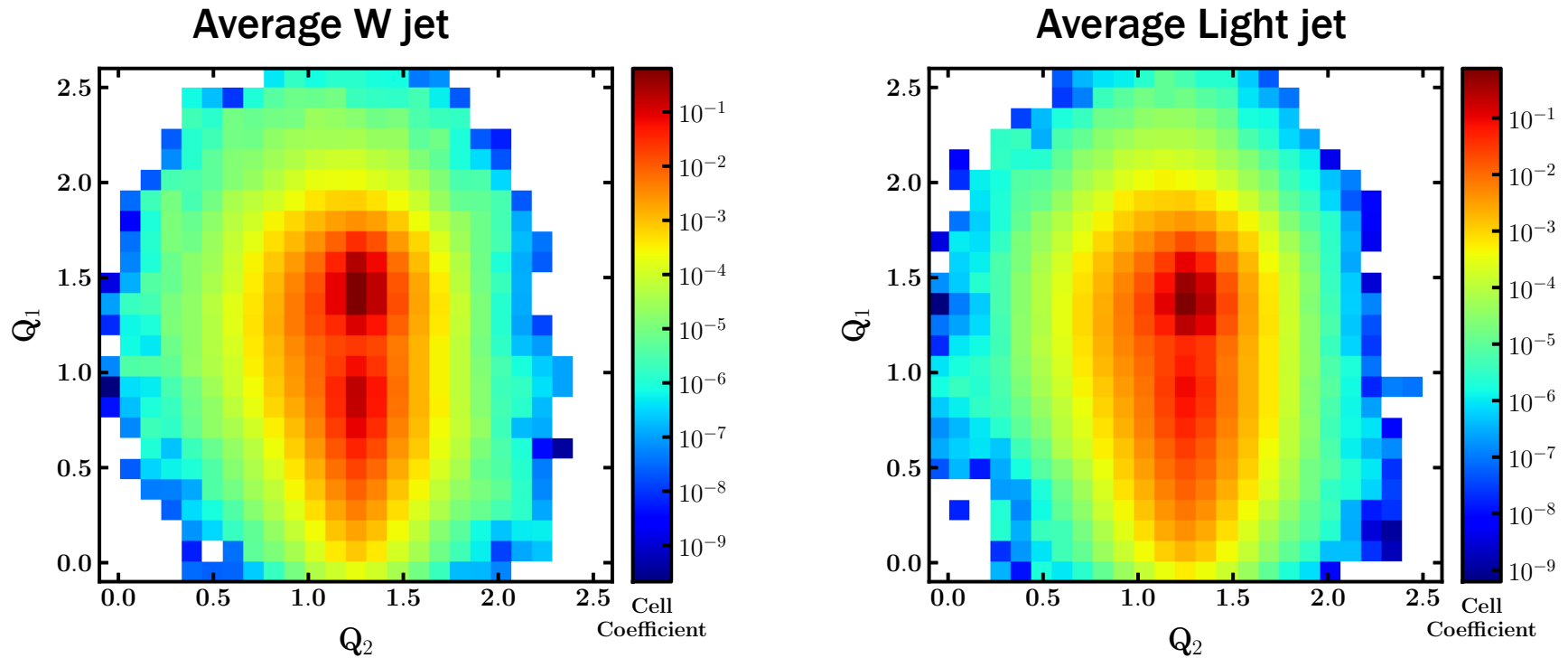
Not much info!

**Average of
rotated W jet**

Much better!



CLASS AVERAGES



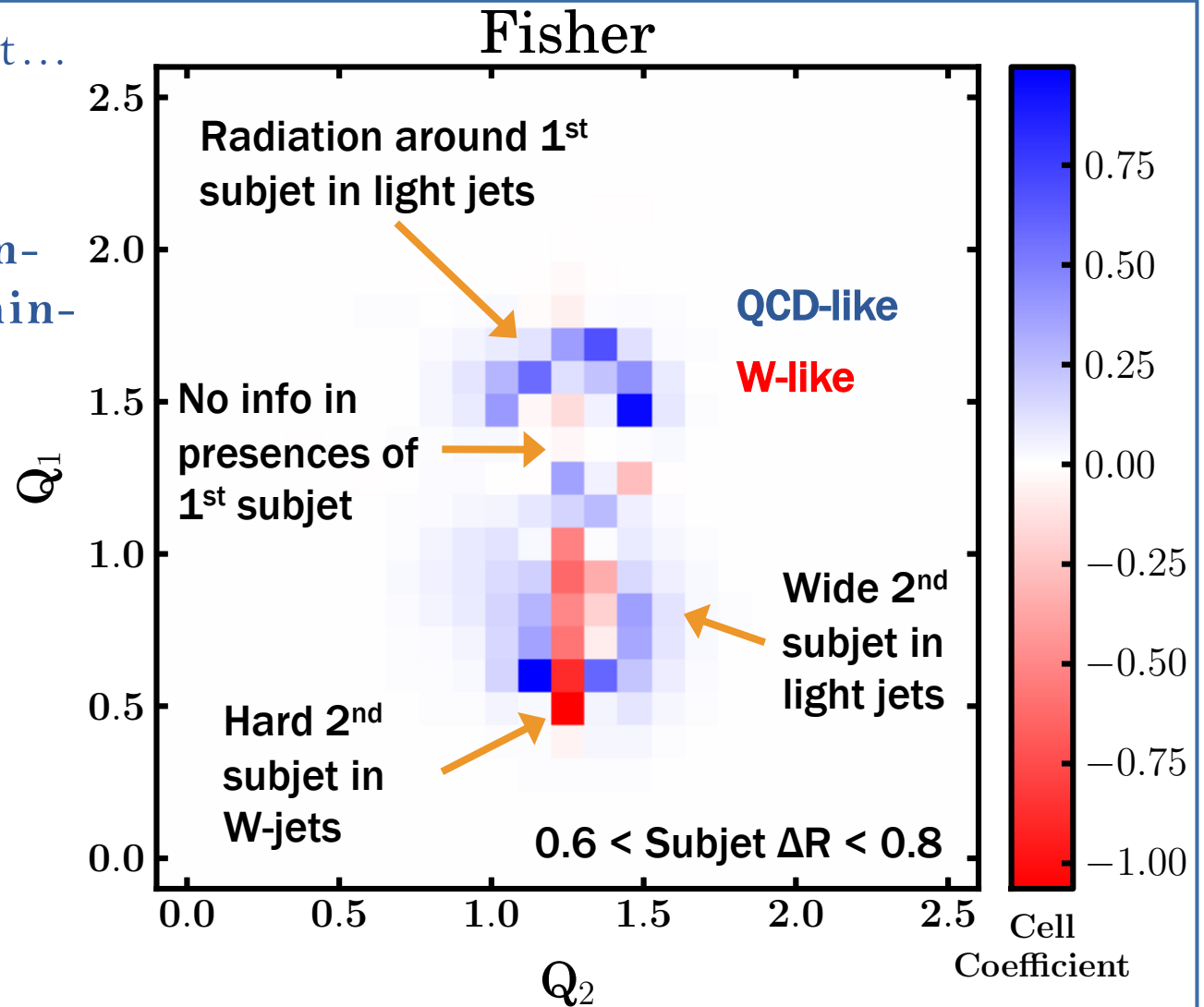
How can we extract the important features?
How can we convert this into discrimination power?

FISHER'S LINEAR DISCRIMINANT

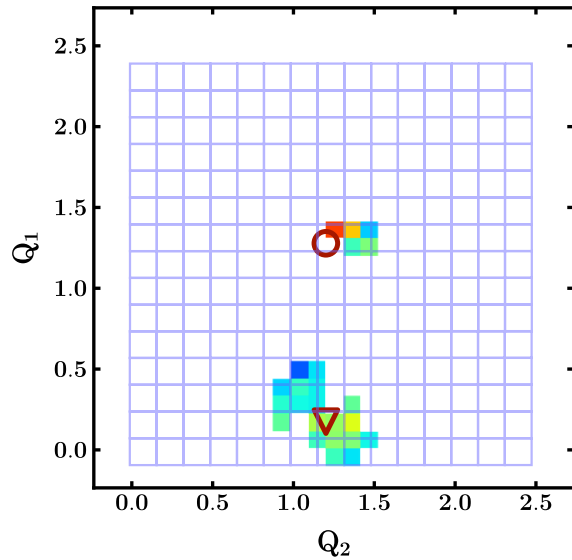
- Finds direction that...
- Maximizes **between-**class scatter / **within-**class scatter
- Extracts the single “most important” feature

FISHER'S LINEAR DISCRIMINANT

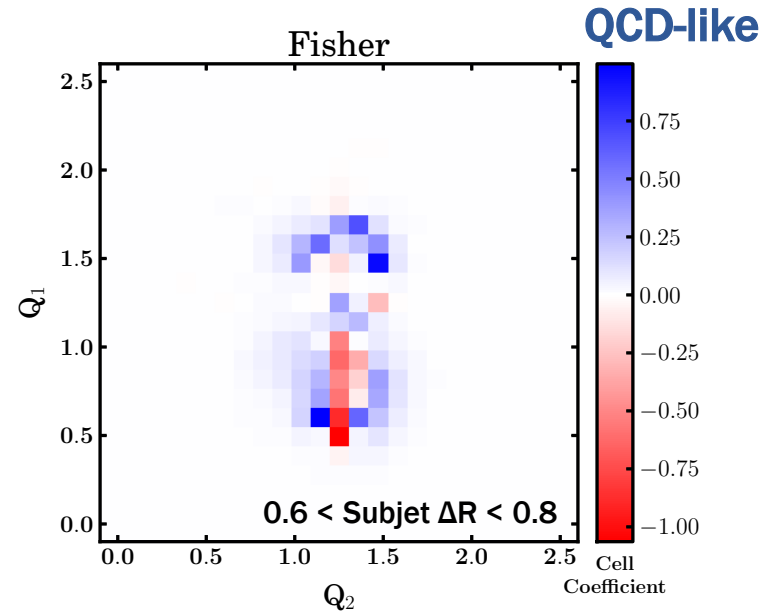
- Finds direction that...
- Maximizes **between-class** scatter / **within-class** scatter
- Extracts the single “most important” feature



EXTRACTING A VALUE



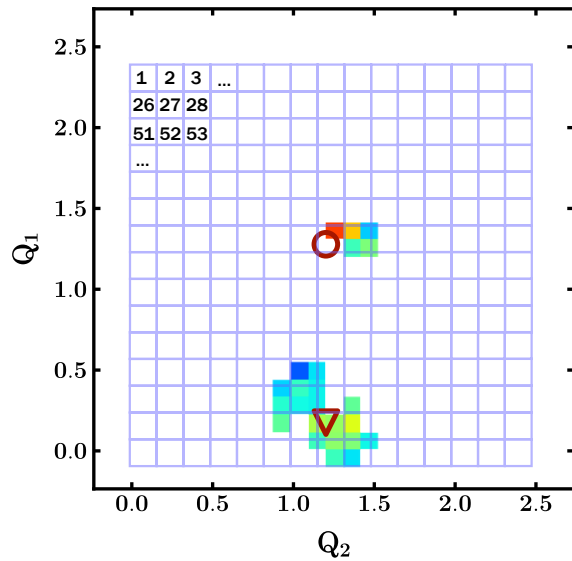
*Jet-image
being tested*



*Jet-image
doing the testing*

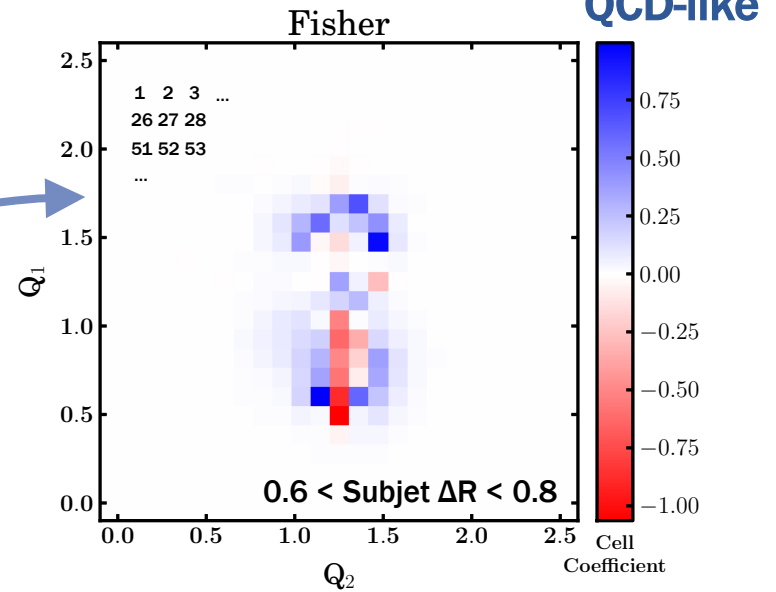
W-like

EXTRACTING A VALUE



*Jet-image
being tested*

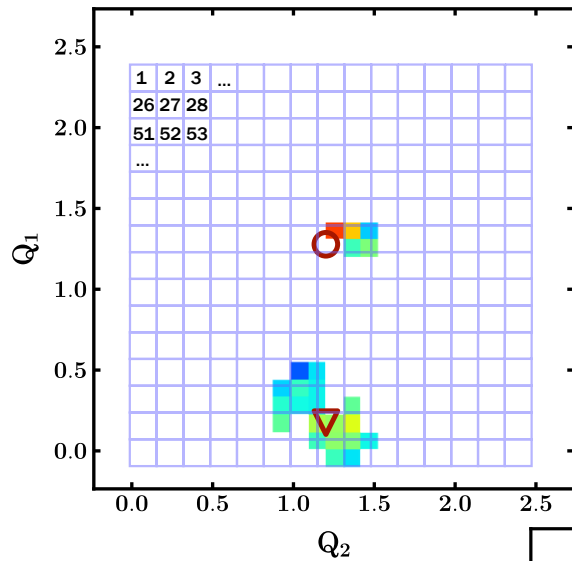
$$D = \sum p_i^T \lambda_i$$



*Jet-image
doing the testing*

W-like

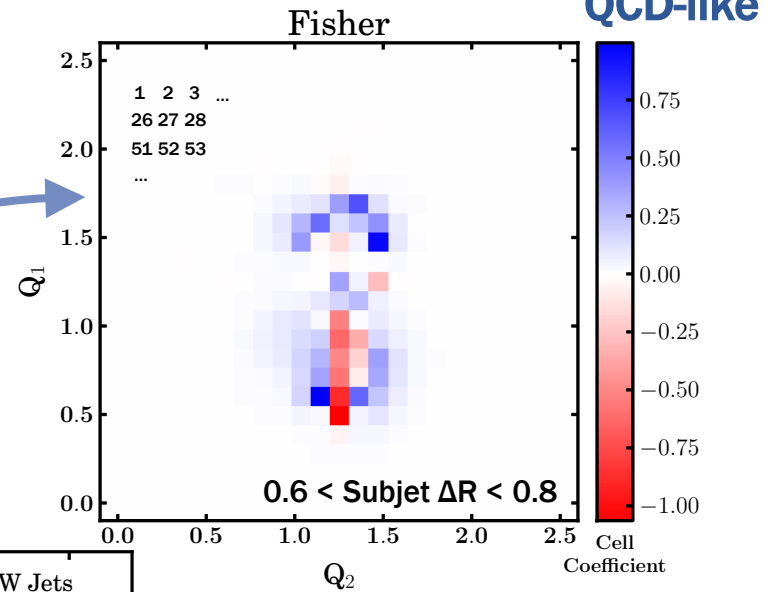
EXTRACTING A VALUE



*Jet-image
being tested*

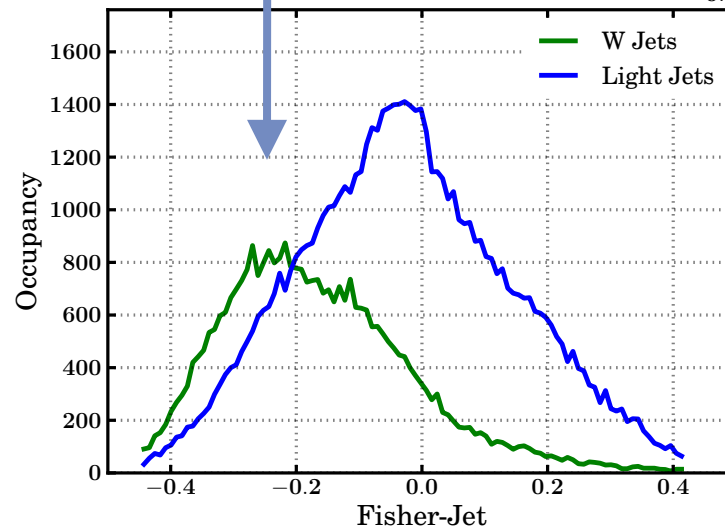
*Dot product of
two jets-images!*

$$D = \sum p_i^T \lambda_i$$



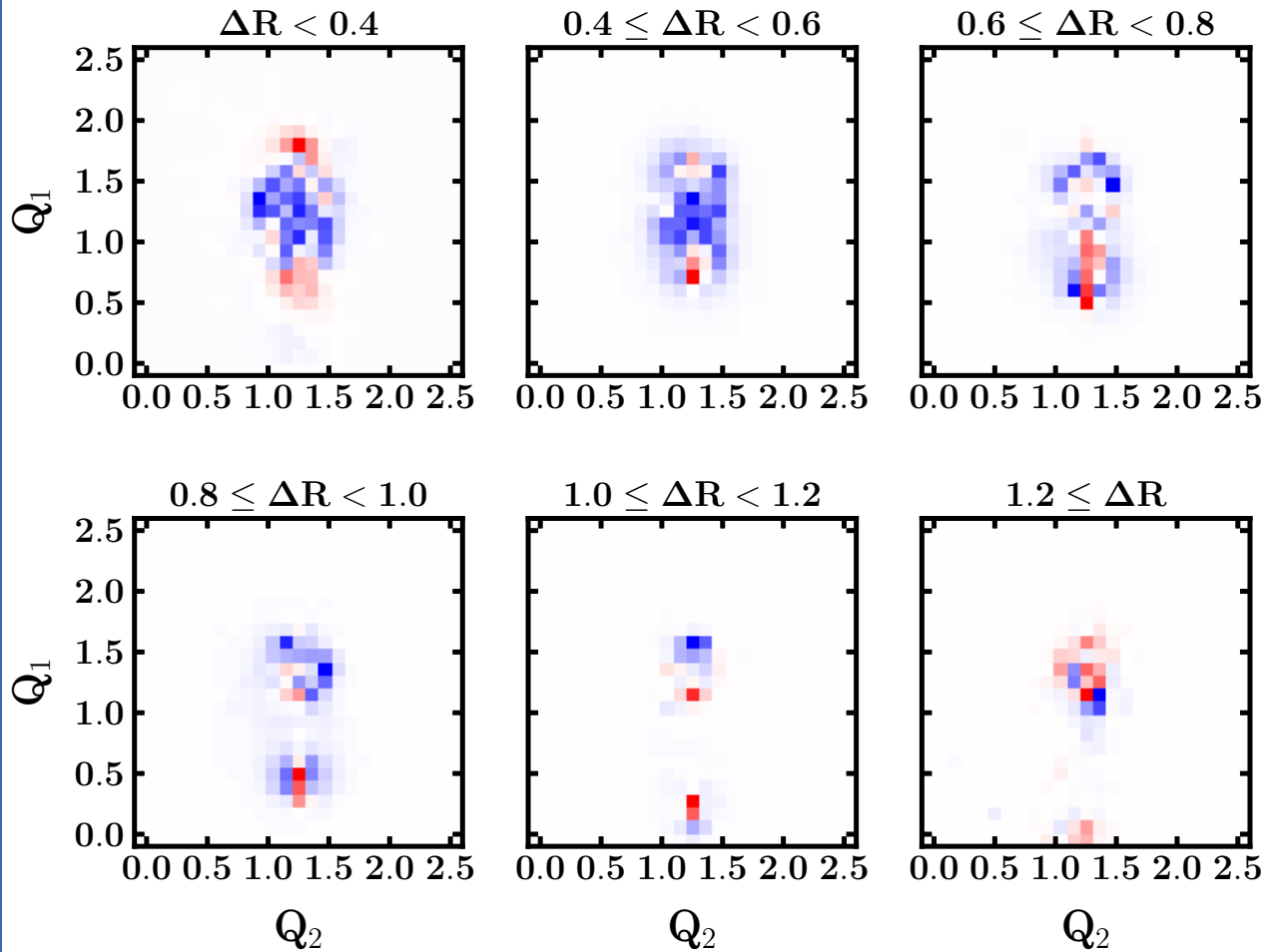
*Jet-image
doing the testing*

W-like



BINNED FISHER

Subjet ΔR -Binned Fisher-Jets **QCD-like** **W-like**

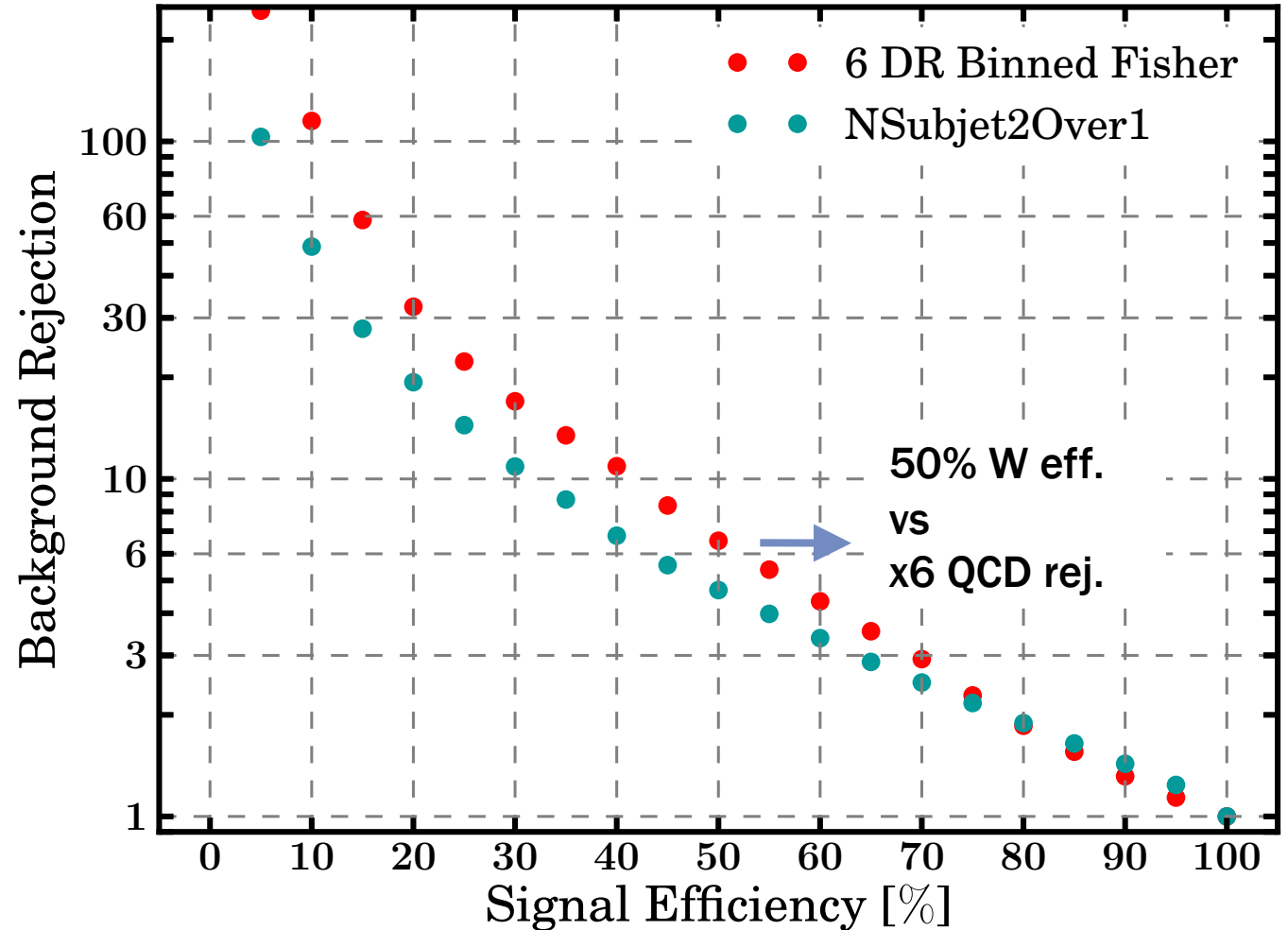


- Hadronic W vs light
- $250 < pT/GeV < 300$
- $55 < Mass/GeV < 105$

Subjet ΔR causes much image variation without too much discrimination

PERFORMANCE

- Use the Fisher' single feature a a discriminant?
- Comparable results to $\beta=1 \tau_2/\tau_1$



DIFFERENT PT BINS

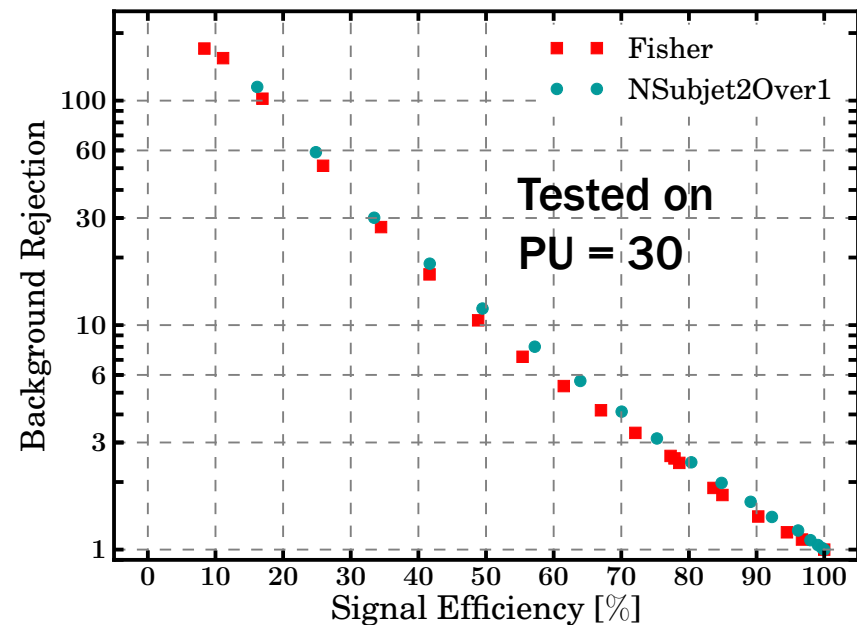
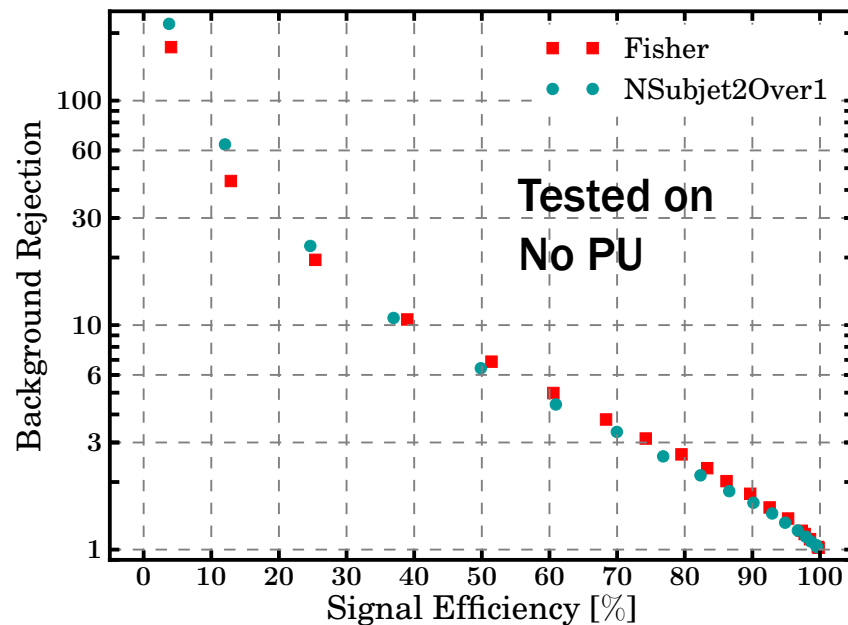
- Train new discriminant for each 50 GeV jet pT window
- More complex binning/image processing could be useful, but performance is robust against jet pT

W Efficiency @ x10 QCD Rejection

Minimum Jet pT	200	250	300	350	400	450
NSubjettiness	29%	32%	34%	34%	39%	40%
6 DR Bin Fish	34%	40%	41%	44%	46%	44%

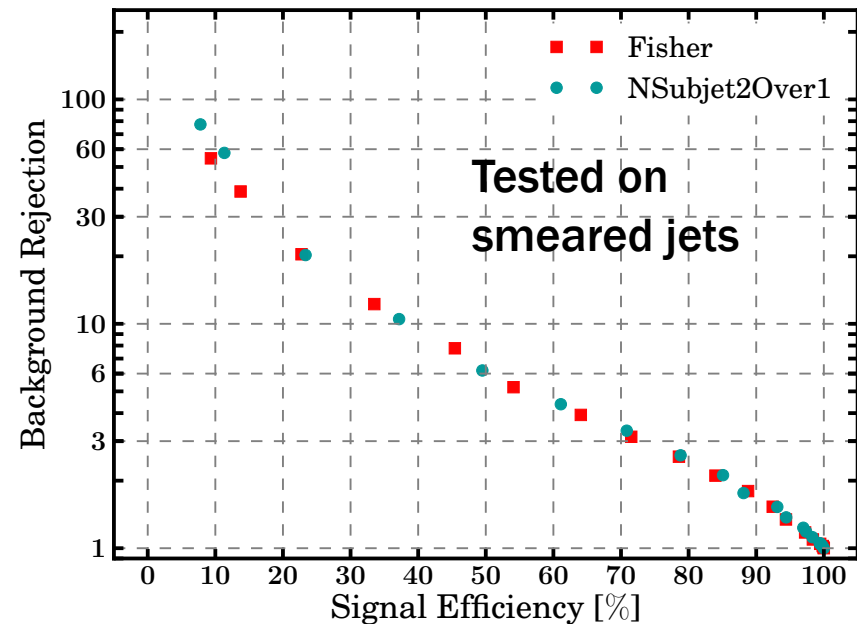
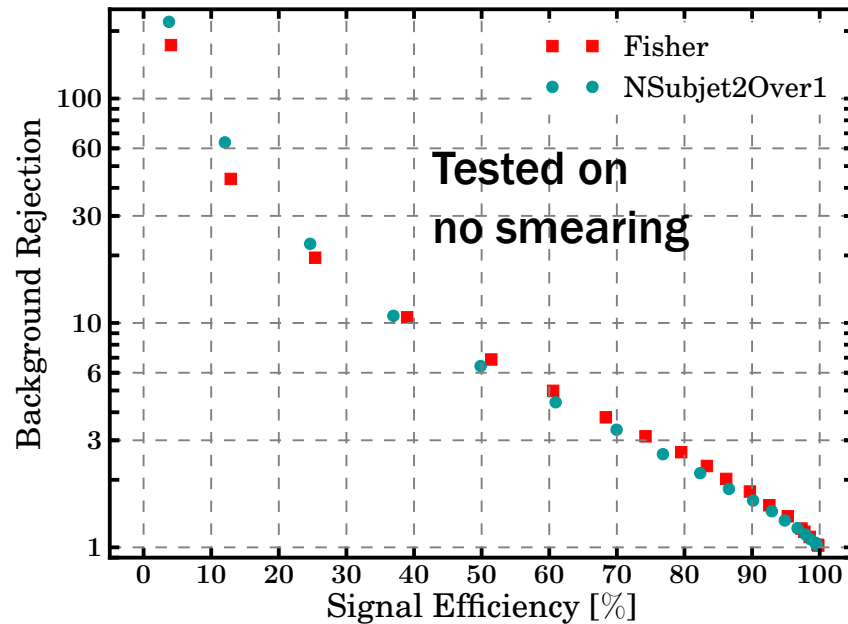
PRELIMINARY CROSS CHECKS

- Train on no PU samples
- Test on samples of W and light jets with $\mu=30$
- Our performance gets “better”
 - Only because we add jets to the bkg sample that are very easy to remove
 - Happens for N-subjettiness too



CROSS CHECK: SMEARING

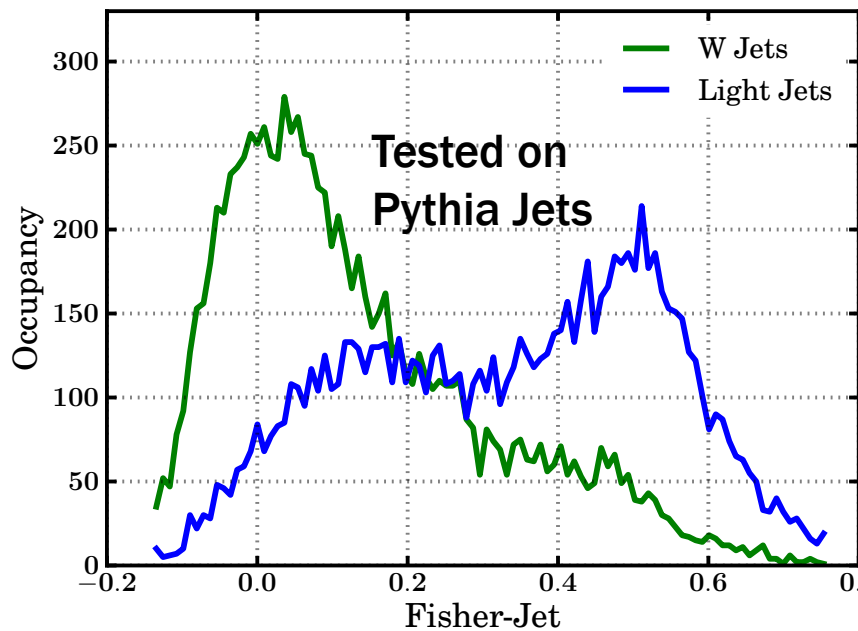
- As a first pass, lets try applying Peter Loch's "smearing"
 - Produces a cell-to-cell smearing by simulating particle in material
 - No significant effect on performance



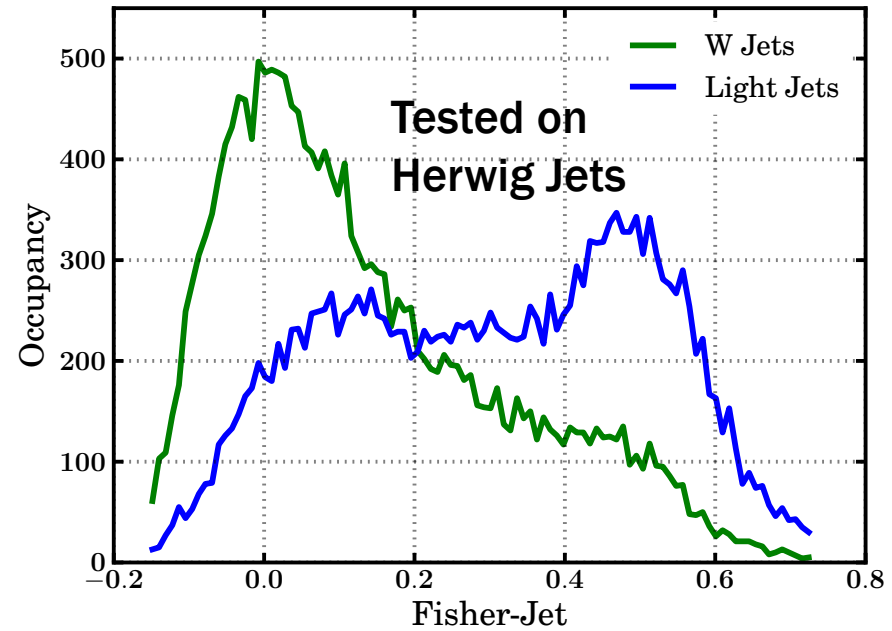
CROSS CHECKS: GENERATORS

- Check Pythia vs Herwig: want to see no sensitivity to differences
- Use the same discriminant (trained on Pythia jets)
- Similar performance when tested on Pythia or Herwig jets

W eff @ QCD Rejection
71% @ x2
18% @ x10
10% @ x20



W eff @ QCD Rejection
70% @ x2
20% @ x10
11% @ x20

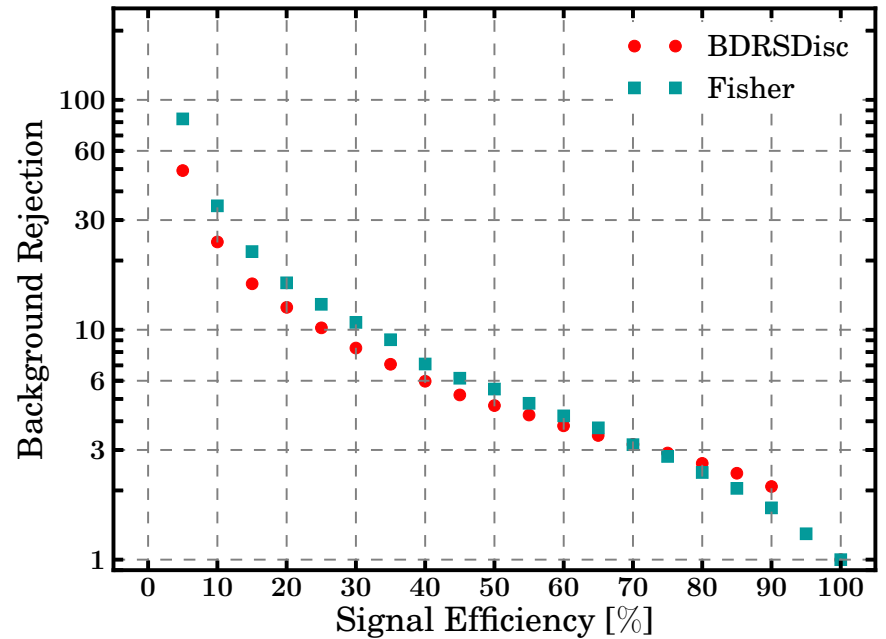
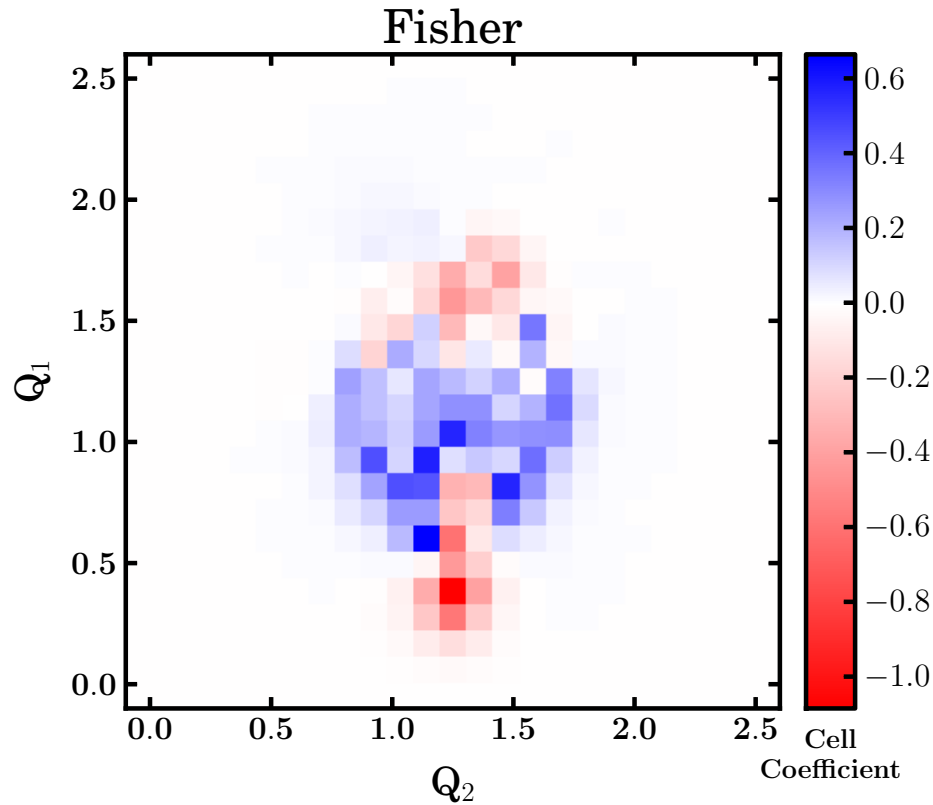


GENERALIZE

Use exact same technology
for many different systems!

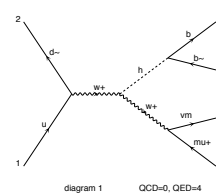
Lets take a *peek...*

HIGGS TO B-QUARKS

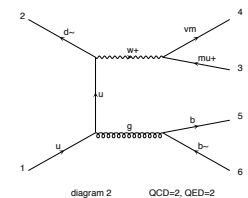
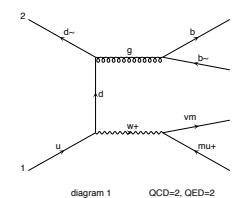


Generate with MG5
Showered with Pythia8
Two G.A. btagged subsets
 $125 > \text{Jet Mass} > 110$

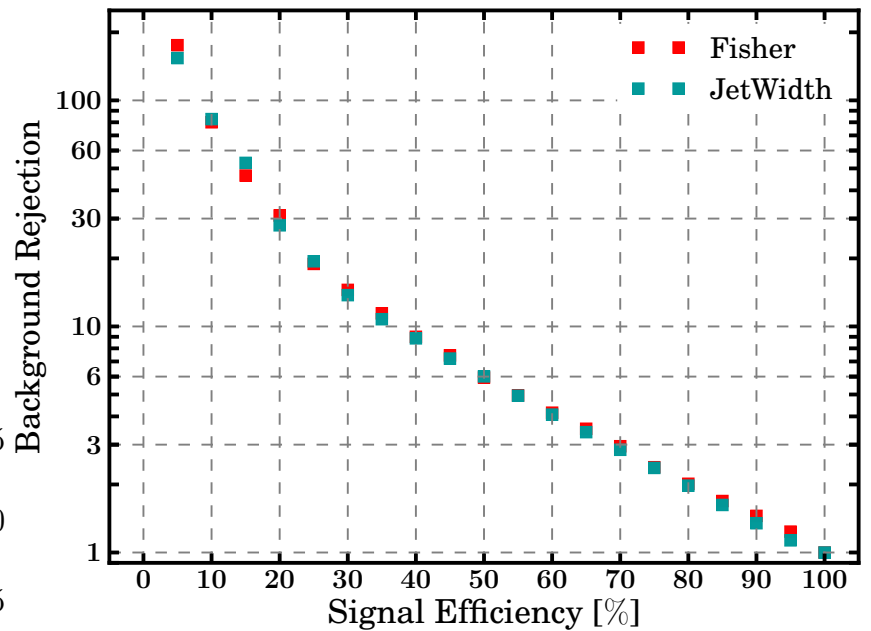
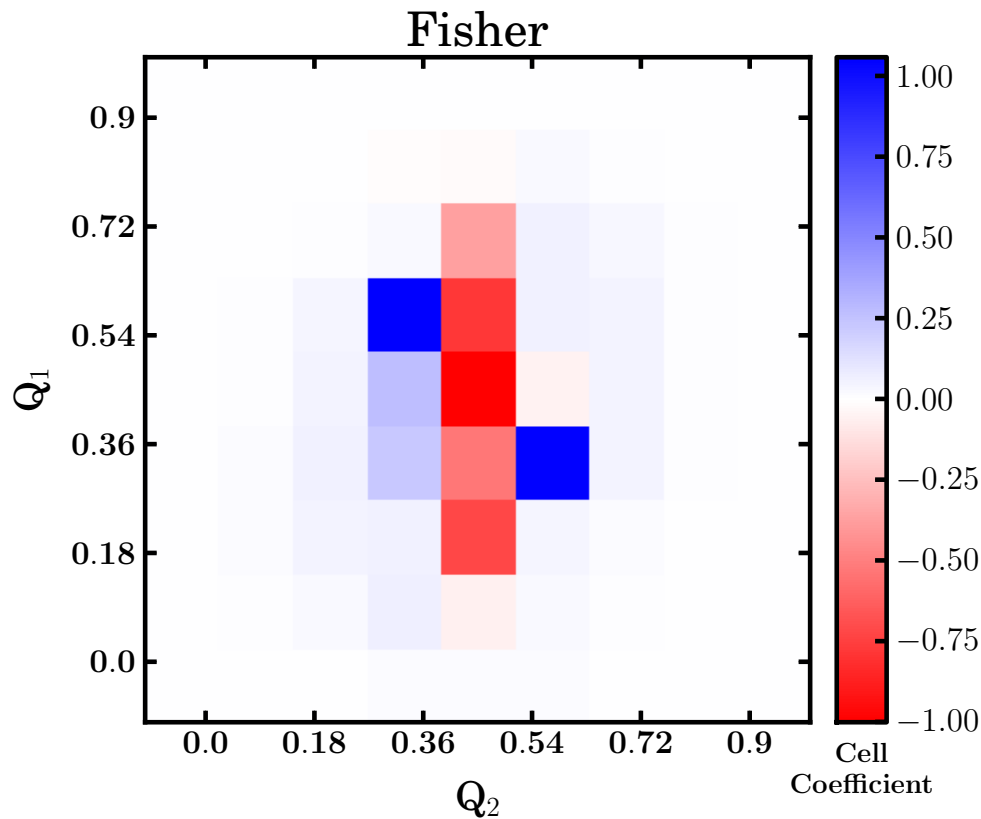
$H \rightarrow bb$



$g \rightarrow bb$



QUARK VS GLUON

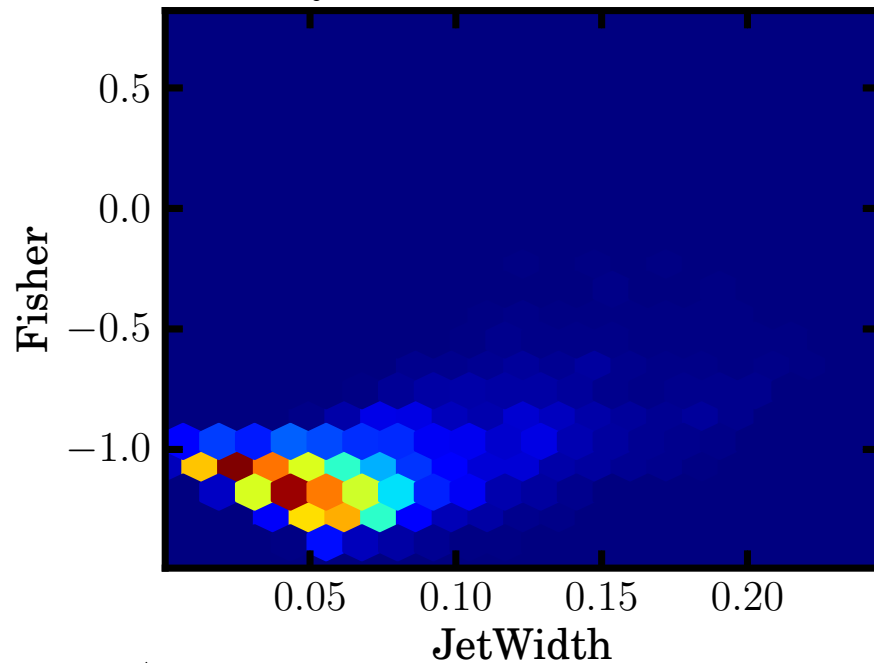


Identical performance to width

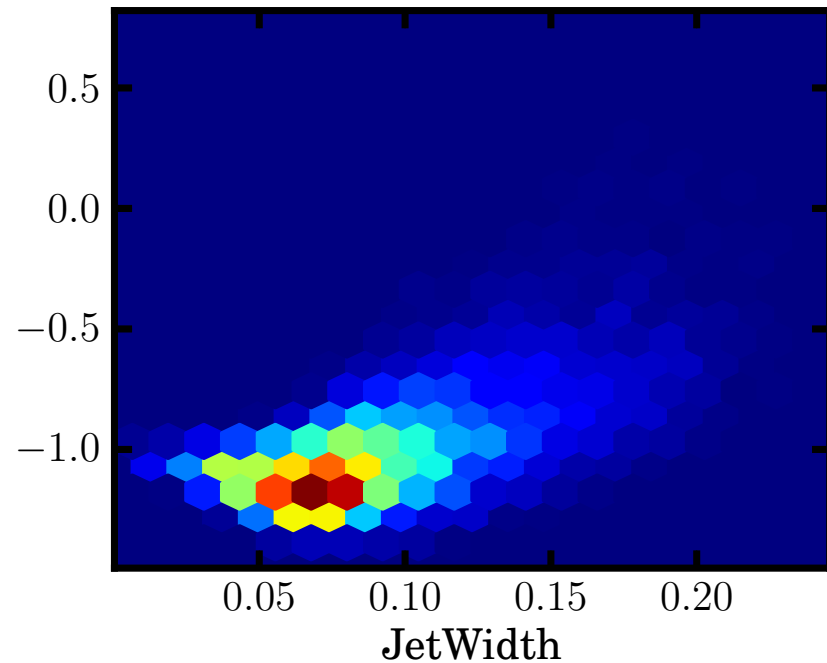
QUARK VS GLUON

Low pT, 0.4, Trimmed Jets: Pythia8, No PU, No smearing

Quark Jets (corr = 0.41)



Gluon Jets (corr = 0.62)



Width and Fisher have the same performance, but aren't completely correlated

FUTURE OPPORTUNITIES

- Many other systems to study
 - Top vs light, hadronic higgs, q/g
 - Hadronic taus, electrons in jets, photon vs π^0
 - Whole calorimeter event-images for high multiplicity signatures
 - Pile up quantification is the perfect linear problem
- Not sure preprocessing is optimized
 - Normalization, center fixing, scaling, boosting, whitening...
 - Starting from other physical bases (eg. cell pT pairs vs cell pTs)
- Today's methods only took us the year 2000 in the CS literature
 - Facial recognition is very active and new methods abound pouring out!
 - More powerful linear methods, non-linear dimensional reduction, etc...

CONCLUSION

- Conceptual mapping between **jets and images** allowed us to exploit well-studied **CS techniques** to jet flavor tagging
- Using easy-to-examine **Fisher's** linear discriminant, produced W vs light **separation at or above N-subjettiness**
 - No need to define 'good' variables *a priori* → **flexible method!**
- Discriminant shows little sensitivity to changing: pile up, detector smearing, event generator

BACK UP

MATH OF FISHER

- Simplification in the case of only 2 classes...

$$[\vec{\mu}_W]_k = \frac{1}{N} \sum_{jets} p_k^T$$

Average W jet

$$[C_W]_{i,k} = \frac{1}{N} \sum_{jets} (p_k^T - \mu_k)(p_i^T - \mu_i)$$

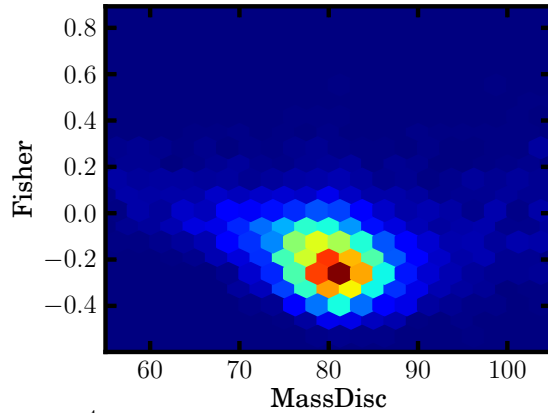
Covariance of cell p^T s in W jets

$$\vec{F} = (C_W + C_L)^{-1} (\vec{\mu}_W - \vec{\mu}_L)$$

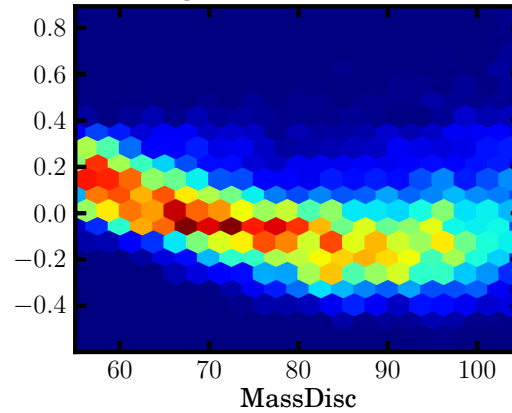
FISHER 2D

250, R=1.2, Trimmed Jets: Pythia8, No PU, No smearing

W Jets (corr = -0.15)



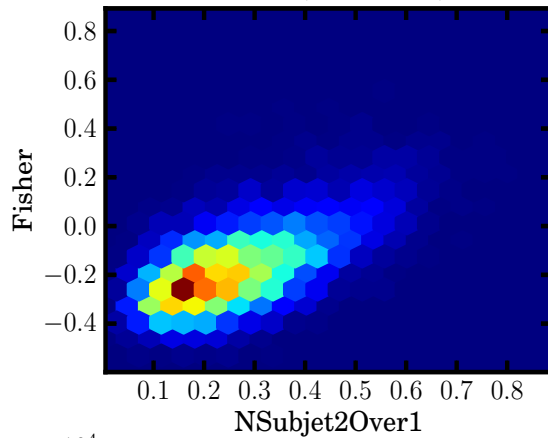
Light Jets (corr = -0.34)



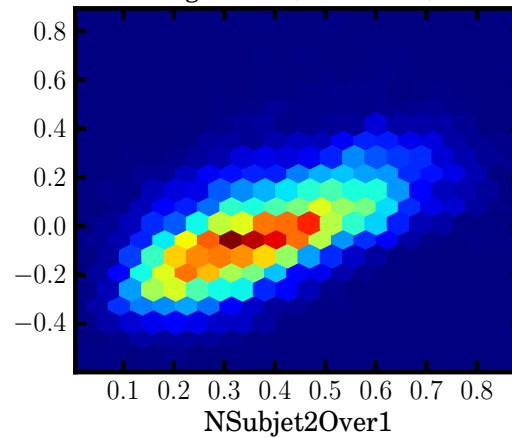
$0.6 < \text{Subjet } \Delta R < 0.8$
 $55 < \text{Mass} < 105$
 $250 < \text{jet pt} < 300$

250, R=1.2, Trimmed Jets: Pythia8, No PU, No smearing

W Jets (corr = 0.60)



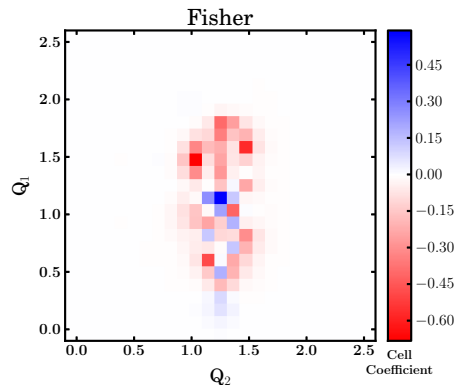
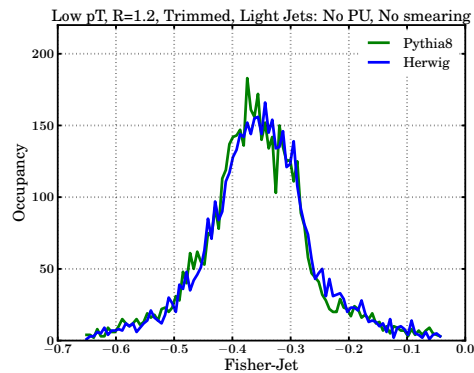
Light Jets (corr = 0.63)



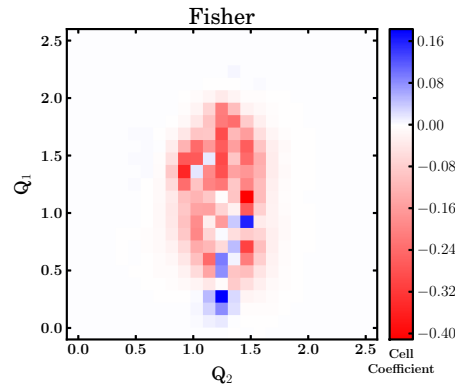
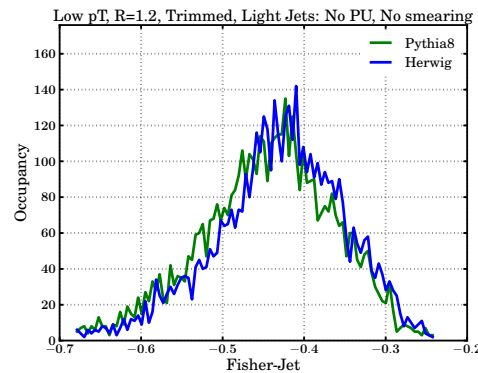
COMPARING GENERATORS

- Is there linear discrimination between Pythia / Herwig?
- Nope! But this is how they differ linearly

W Jets



Light Jets



Train on Pythia **VS** Herwig

- There are some small but measurable differences
- Further study needed but I'd hypothesize its about the pT balance

Pythia-like
Herwig-like