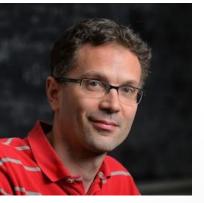
Generalization bounds for uniformly stable algorithms

Vitaly Feldman Google Brain

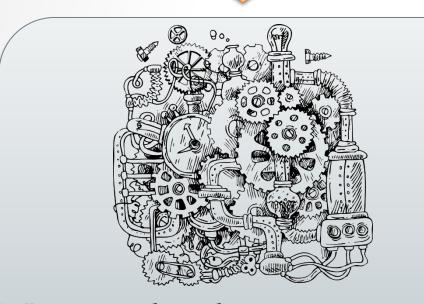
with Jan Vondrak





Generalization bounds

Dataset $S = (z_1, \dots, z_n) \sim P^n$



Data distribution P Loss function $\ell(w, z)$ Generalization error/gap for w = A(S): $\Delta_S(\ell(w)) = \mathbf{E}_{z \sim P}[\ell(w, z)] - \frac{1}{n} \sum_{i=1}^n \ell(w, z_i)$

Learning algorithm A



Uniform stability [Bousquet,Elisseeff '02]

A has uniform stability γ w.r.t. loss ℓ if for all S, S' that differ in a single element and $z \in Z$ $|\ell(A(S), z) - \ell(A(S'), z)| \leq \gamma$

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Examples:

- Strongly convex ERM [BE '02; Shalev-Shwartz, Shamir, Srebro, Sridharan '09]
- Gradient descent on convex smooth losses [Hardt, Recht, Singer '16]

Typical $\gamma = 1/\sqrt{n}$

From stability to generalization

For ℓ with range [0,1] and A with uniform stability $\gamma \in \left[\frac{1}{n}, 1\right]$ [Rogers,Wagner '78]

$$\mathbf{E}_{S\sim P^n}[\Delta_S(\ell(A))] \leq \gamma$$

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[Bousquet,Elisseeff '02]

$$\Pr_{S \sim P^n} \left[\Delta_S(\ell(A)) \ge \gamma \sqrt{n} \log(1/\delta) \right] \le \delta$$

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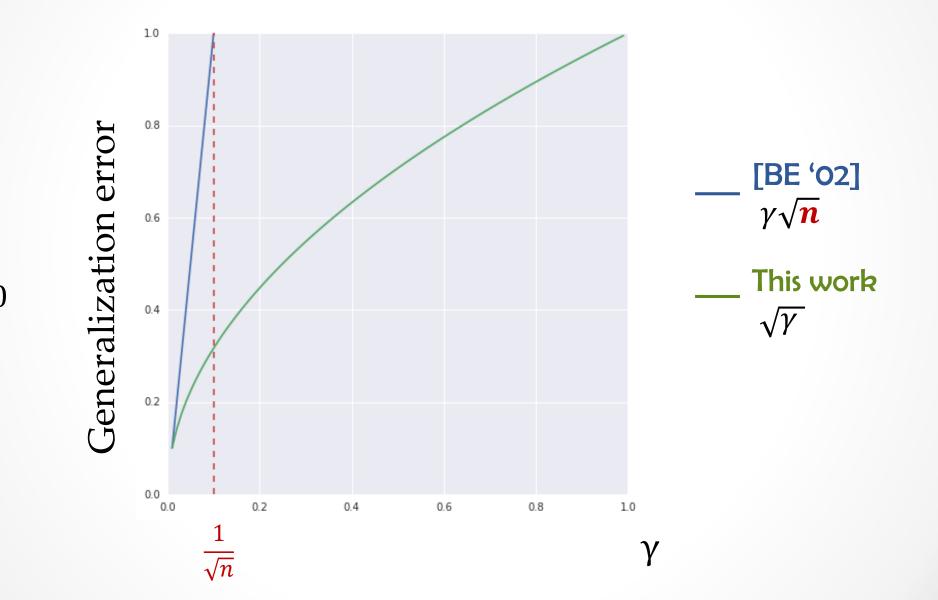
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NEW
$$\Pr_{S \sim P^n} \left[\Delta_S(\ell(A)) \ge \sqrt{\gamma \log(1/\delta)} \right] \le \delta$$

Comparison



n = 100

Second moment

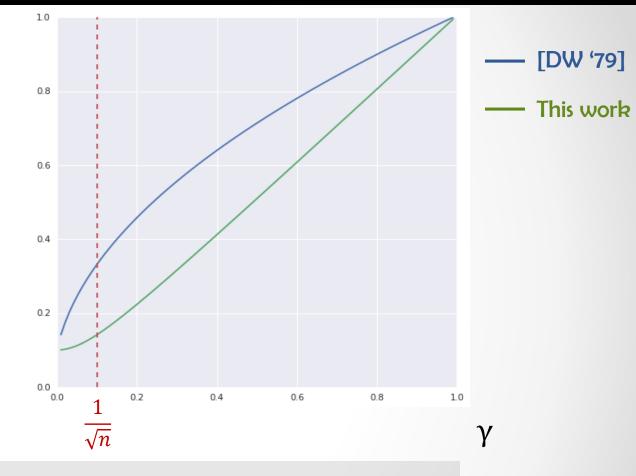
[Devroye,Wagner '79; BE '02]

$$\sqrt{\mathop{\mathbf{E}}_{S\sim P^n} \left[\Delta_S \left(\ell(A) \right)^2 \right]} \leq \sqrt{\gamma}$$

Second moment

[Devroye,Wagner '79; BE '02]

$$\sqrt{\mathop{\mathbf{E}}_{S\sim P^n} \left[\Delta_S \left(\ell(A) \right)^2 \right]} \leq \sqrt{\gamma}$$



TIGHT!

$$\frac{\mathbf{E}}{\sum_{S \sim P^n} \left[\Delta_S \left(\ell(A) \right)^2 \right]} \le \gamma + \frac{1}{\sqrt{n}}$$

There is more

- New proof technique
- Applications to stochastic convex optimization
- Connections to learning with differential privacy

