

Progress on the $K_L \rightarrow \pi^+ \pi^- \gamma$ Analysis

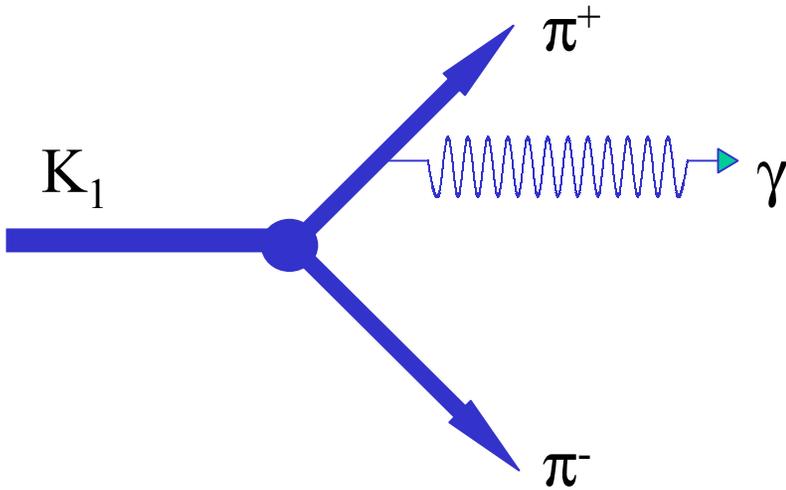
John Shields

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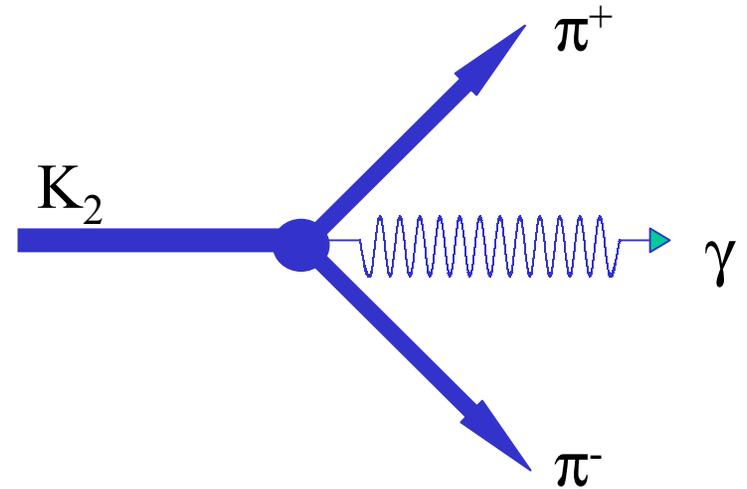
KTeV Collaboration

August 2, 2003

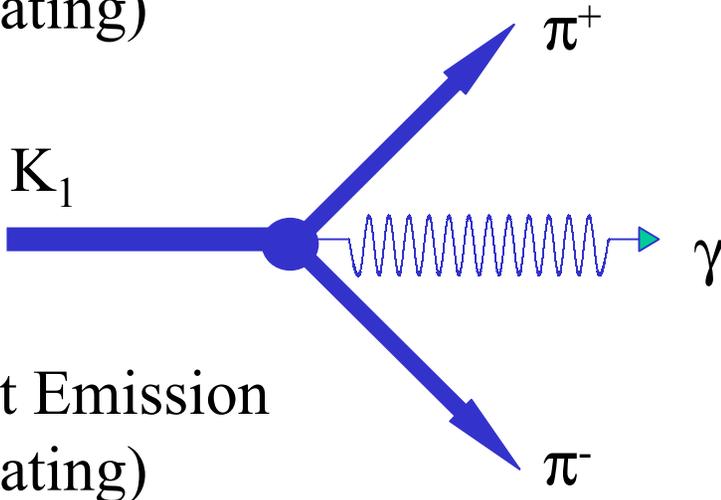
$K_L \rightarrow \pi^+ \pi^- \gamma$ Decay



Inner Bremsstrahlung
(CP Violating)



M1 Direct Emission
(CP conserving)



E1 Direct Emission
(CP Violating)

K_L → π⁺π⁻γ Differential Decay Rate, Sehgal Model

$$\frac{d\Gamma}{d\omega d\cos\theta} = (\text{const}) \left[|\xi_E|^2 + |\xi_M|^2 \right] \omega^3 \beta^3 \left(1 - \frac{2\omega}{M_K} \right) \sin^2\theta$$

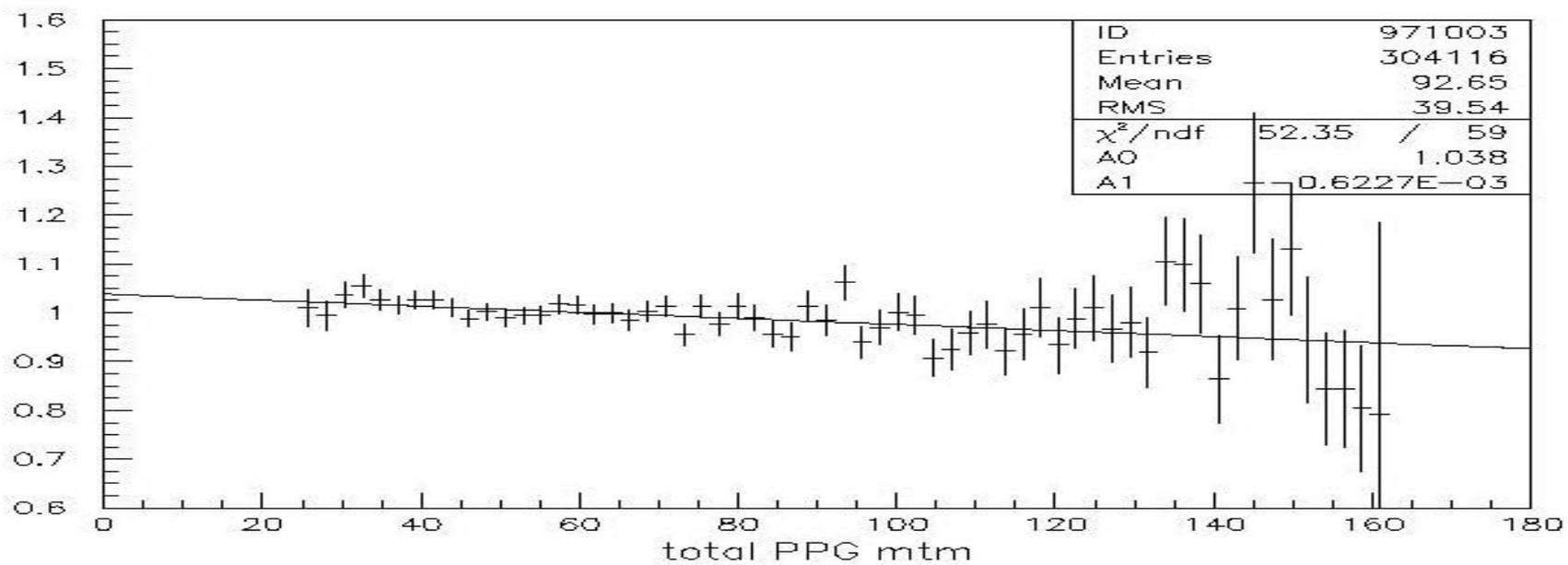
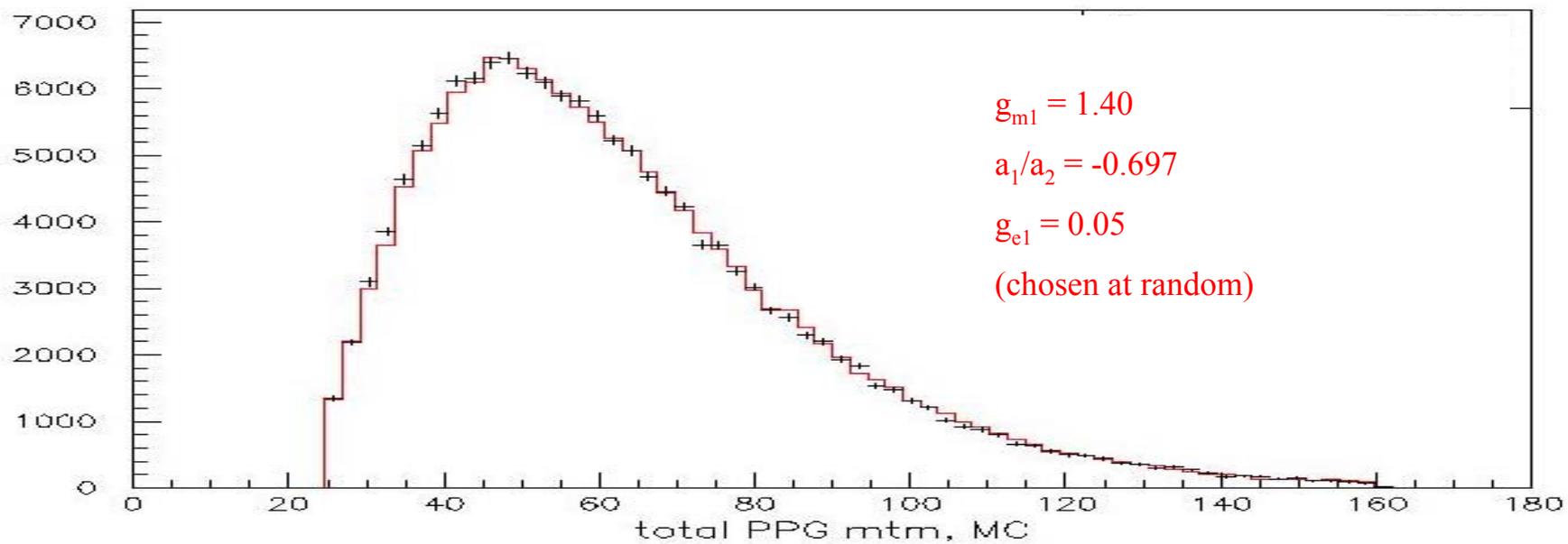
where:

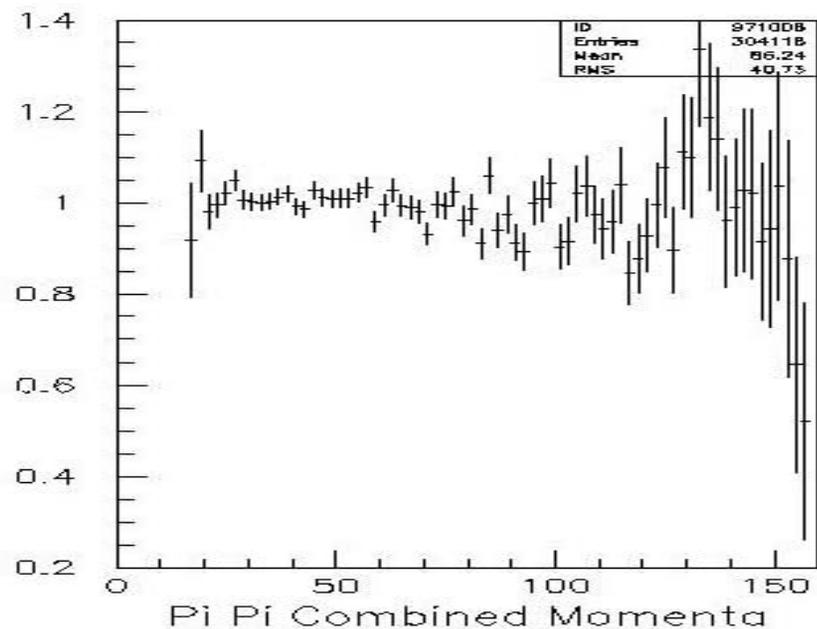
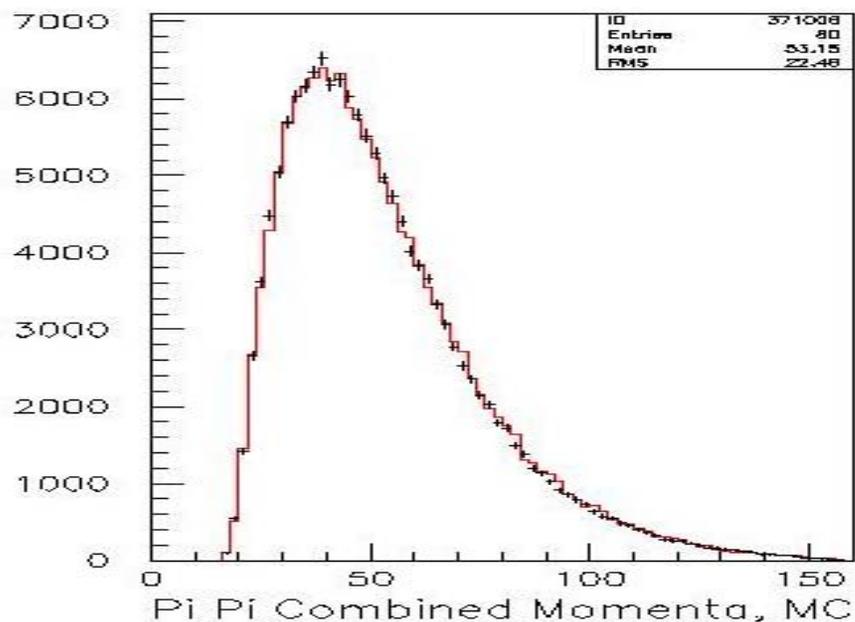
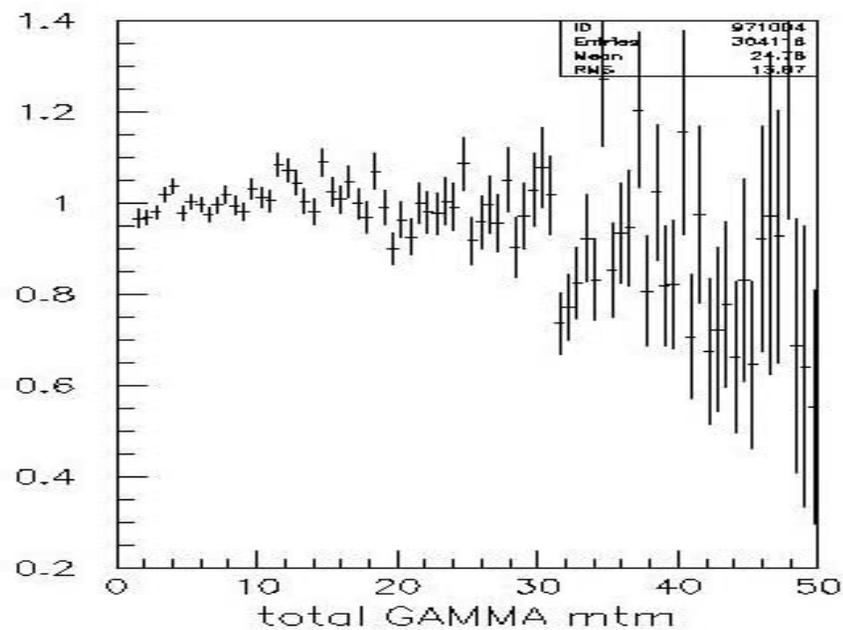
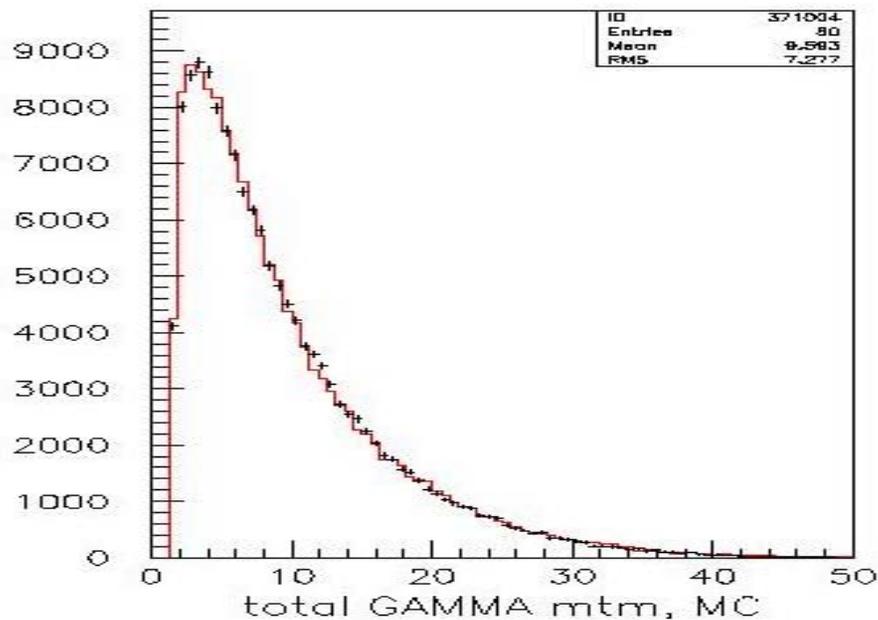
$$|\xi_E| = C_2 \left| \frac{\eta_{+-}}{\omega^2 (1 - \beta^2 \cos^2\theta)} \right| + g_{e1}$$

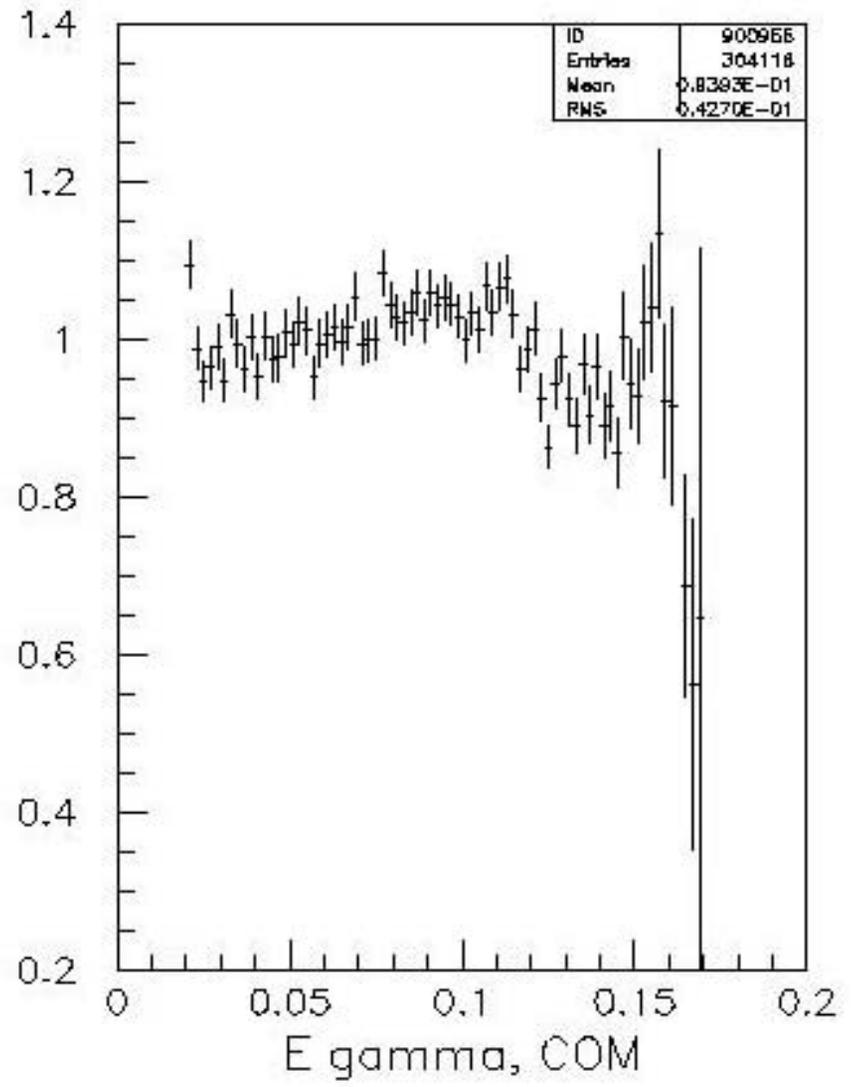
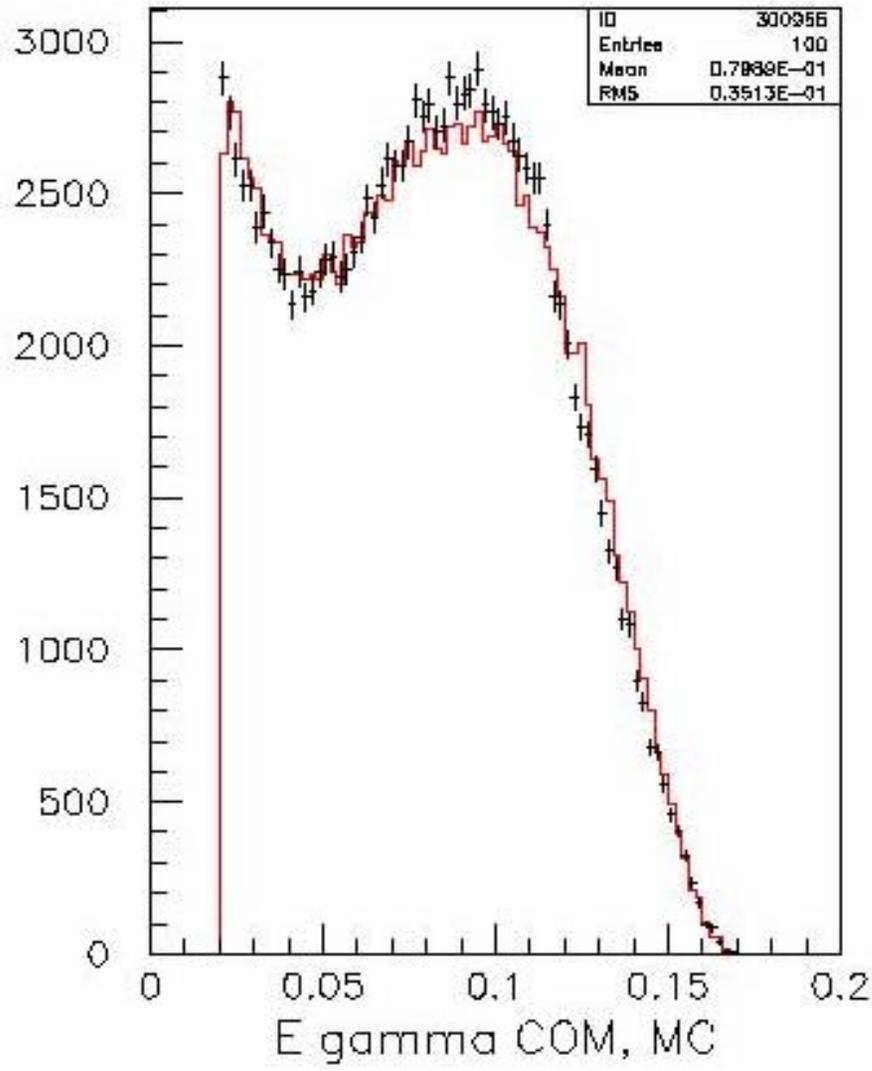
$$|\xi_M| = g_{m1} \left[1 + \frac{a_1/a_2}{\left(M^2_\rho - M^2_K \right) + 2M_K E^*_\gamma} \right]$$

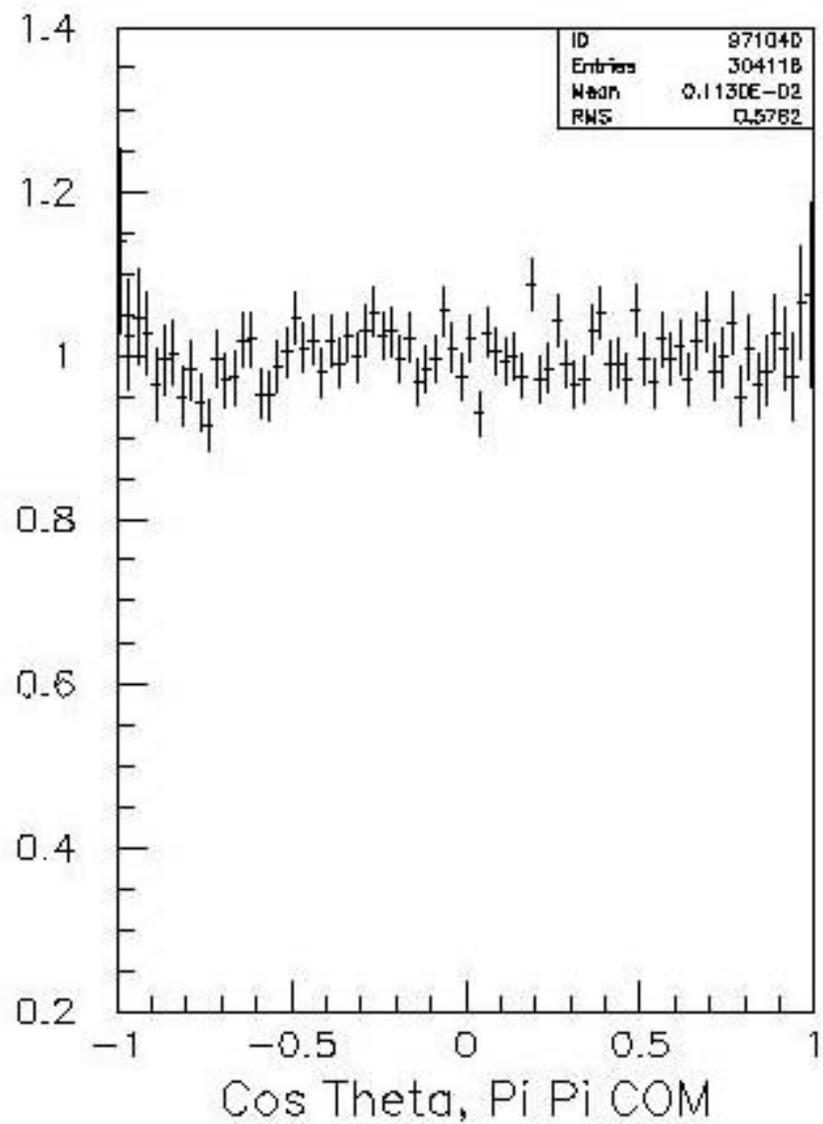
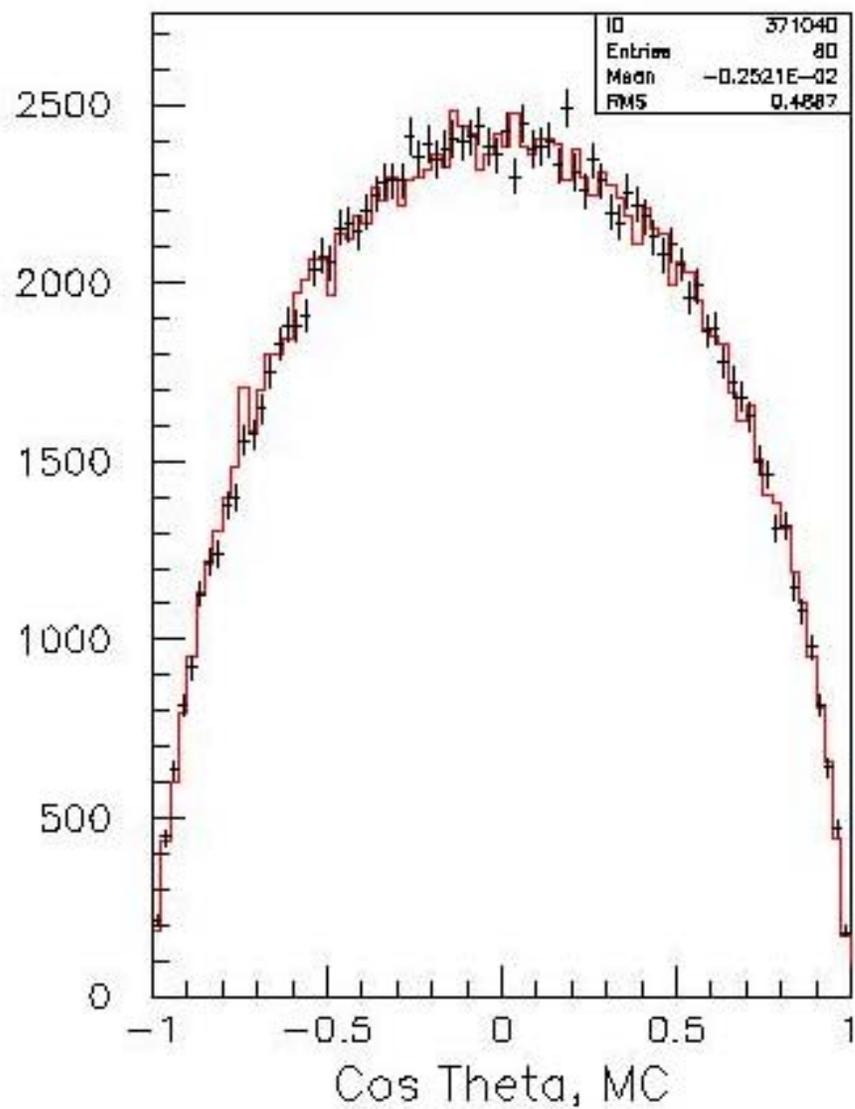
Tasks Completed since Jan, 2003 meeting

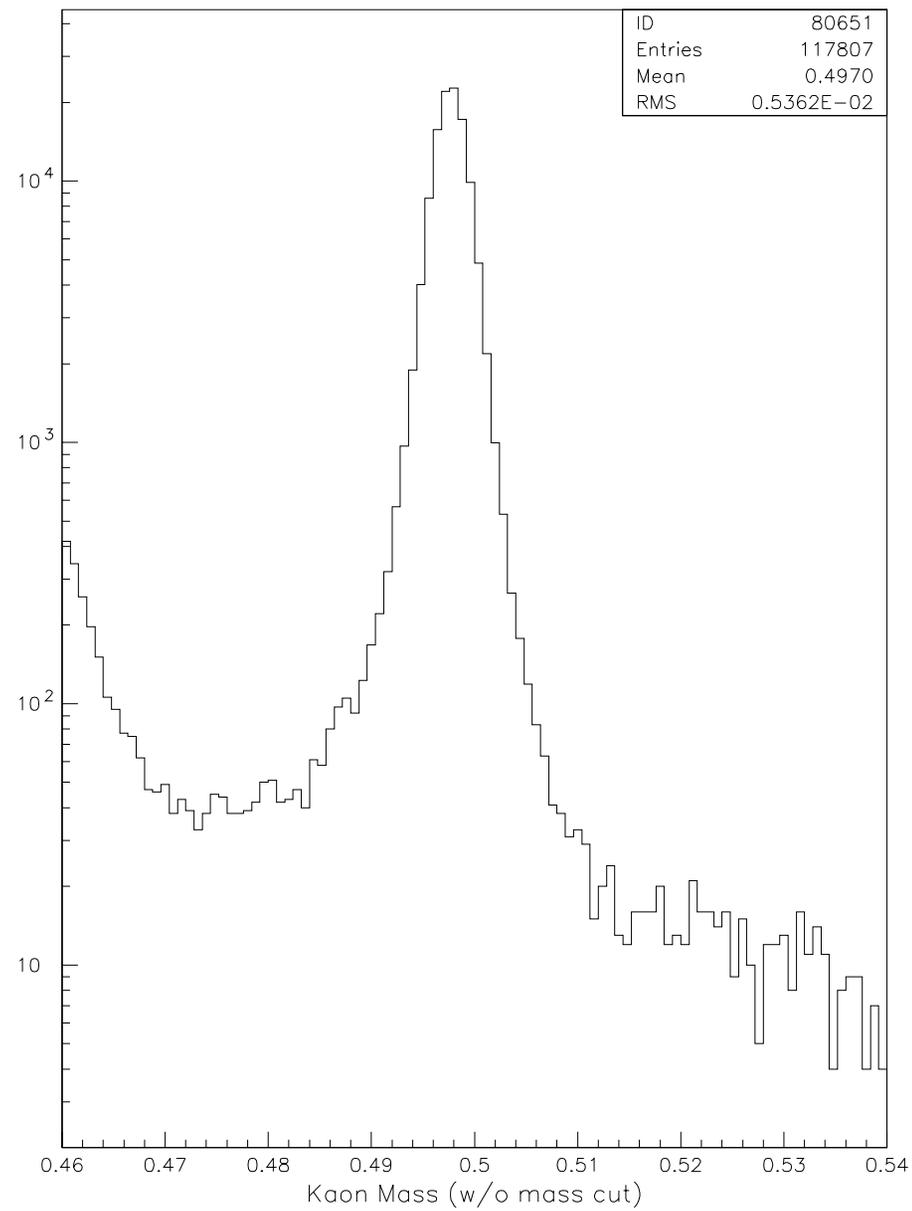
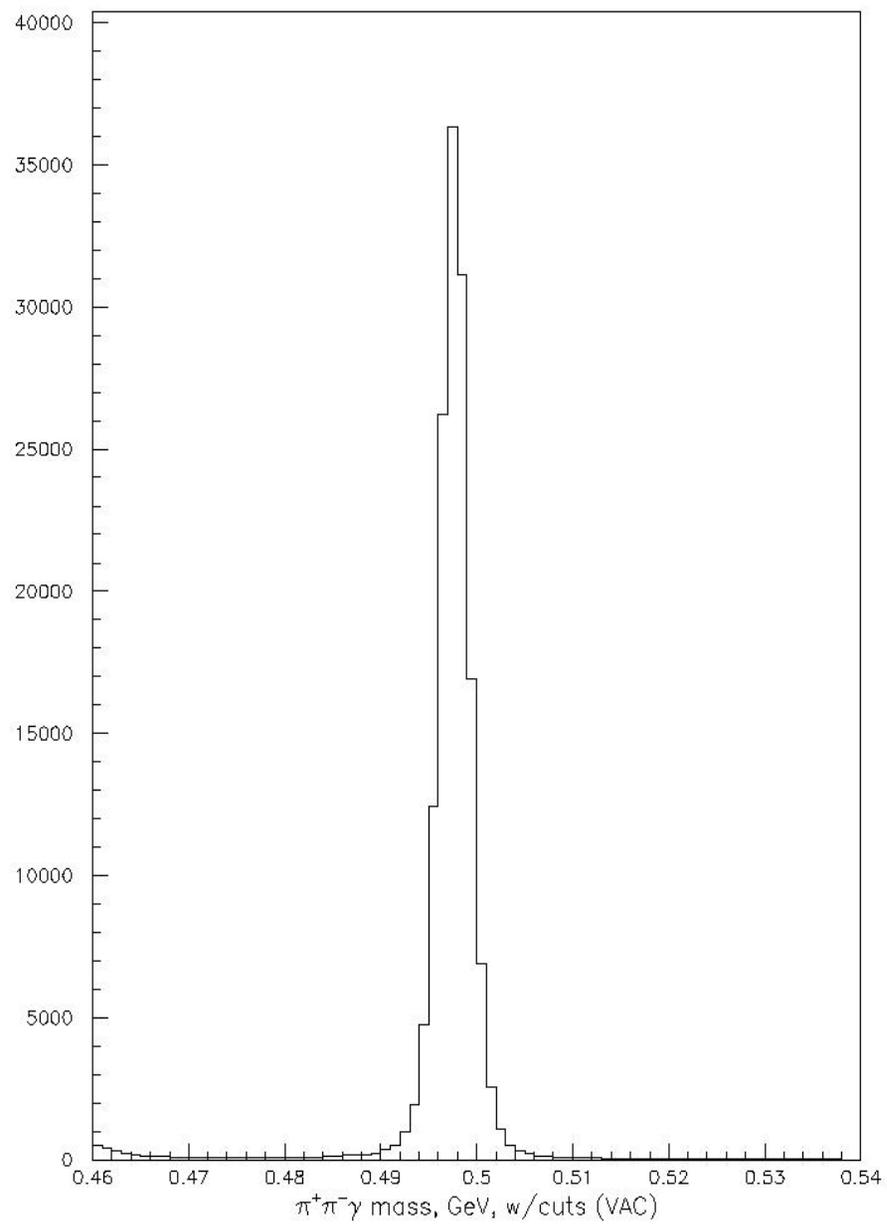
- Upgraded MC and analysis code from Ktevana version v6_00 to v6_01
- Applied RK's L3 filter code correction to the Monte Carlo
 - (accounts for a change in B02 trigger PP0KINE cut during 97 run)
 - (see RK's Jan 3, 2003 write-up for details)
- Added E1 contribution to Monte Carlo matrix element
- Finished implementing a new, more CPU-friendly version of likelihood fitting code **and** tested it extensively on data-sized Monte Carlos with known parameter values











Current Projects

- **Statistical Error Analysis:**

→ determining the size of “Big Monte Carlo” (BMC) sample required to make $\sigma_{\text{BMC}} \ll \sigma_{\text{Ndata}}$ where:

BMC \equiv Big MC used by the likelihood fitting code for reweighting

σ_{BMC} \equiv STAT error due to BMC sample size (only)

σ_{Ndata} \equiv STAT error due to DATA size (only)

– To do this, σ_{BMC} must be determined as a function of “Q”,

where $Q \equiv N_{\text{BMC}} / N_{\text{DATA}}$

Current Projects (cont.)

- So, in other words, we need to know what value of $Q = N_{\text{data}}/N_{\text{BMC}}$ results in $\sigma_{\text{BMC}} \ll \sigma_{N_{\text{data}}}$? ($Q=10?$, $Q=100?$, $Q=1000??$)

- In Theory:

$$\sigma_{\text{BMC}}(Q) = \frac{\sigma_{\text{BMC}}(Q=1)}{\sqrt{Q}}$$

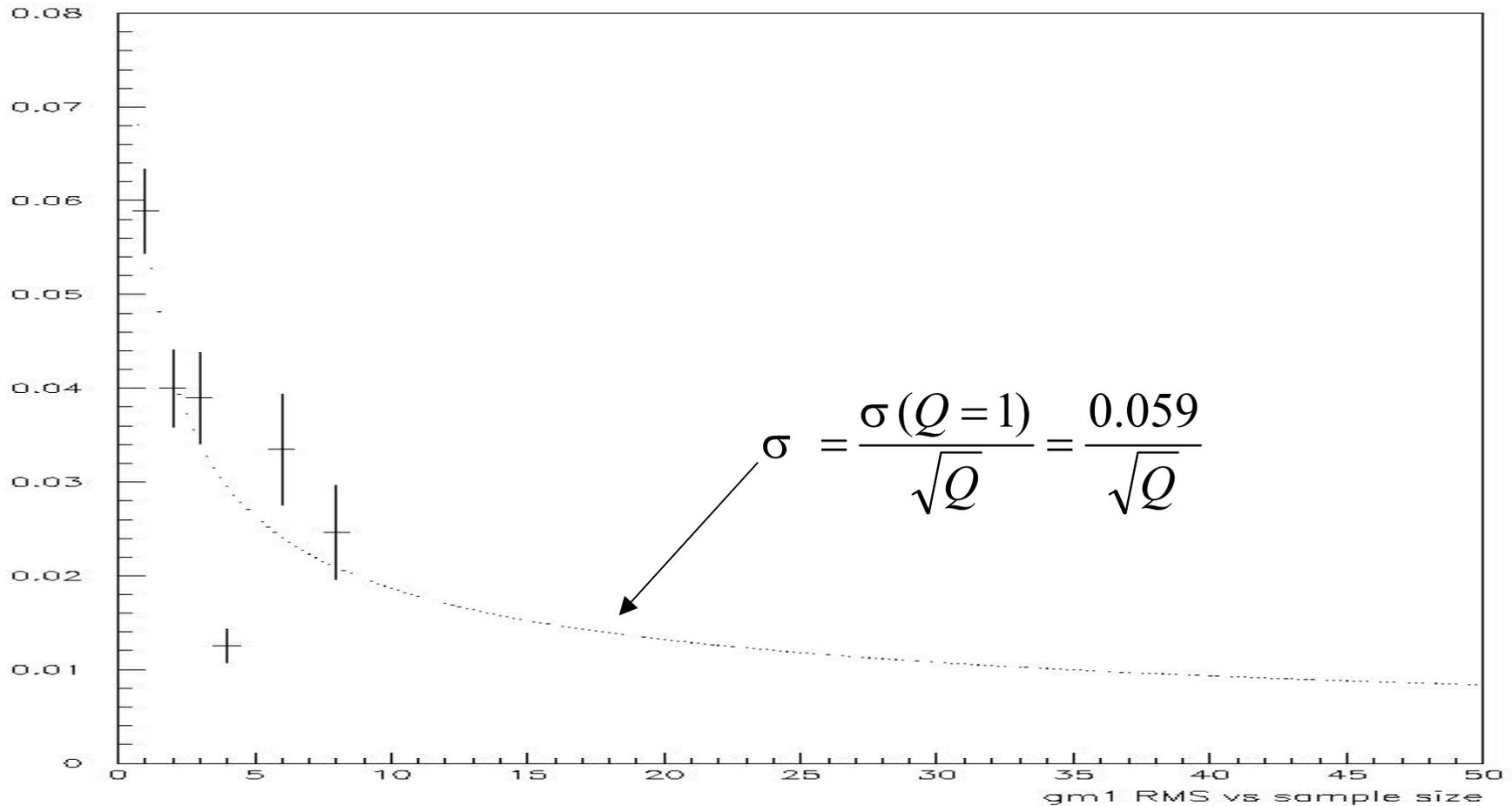
- In Practice: preliminary studies do indeed very *roughly* indicate

$\frac{1}{\sqrt{Q}}$ dependence

– Caveats:

- Studies conducted with very limited number of MC samples (and hence over a very small range of “Q”)
- Further studies with a MUCH greater number of data-sized samples is required to reliably extend the curve for $Q \gg 1$

σ_{BMC} vs. Q (for g_{m1} parameter and limited samples)



PLOT: Distribution of likelihood fits of the SAME “data” sample by different Q=2 “Big Monte Carlos”

• Why are we so concerned about the exact Shape of the σ_{BMC} curve?

• likelihood fit process is massively CPU-intensive (since Ndata \sim 120K) and scales linearly in Q

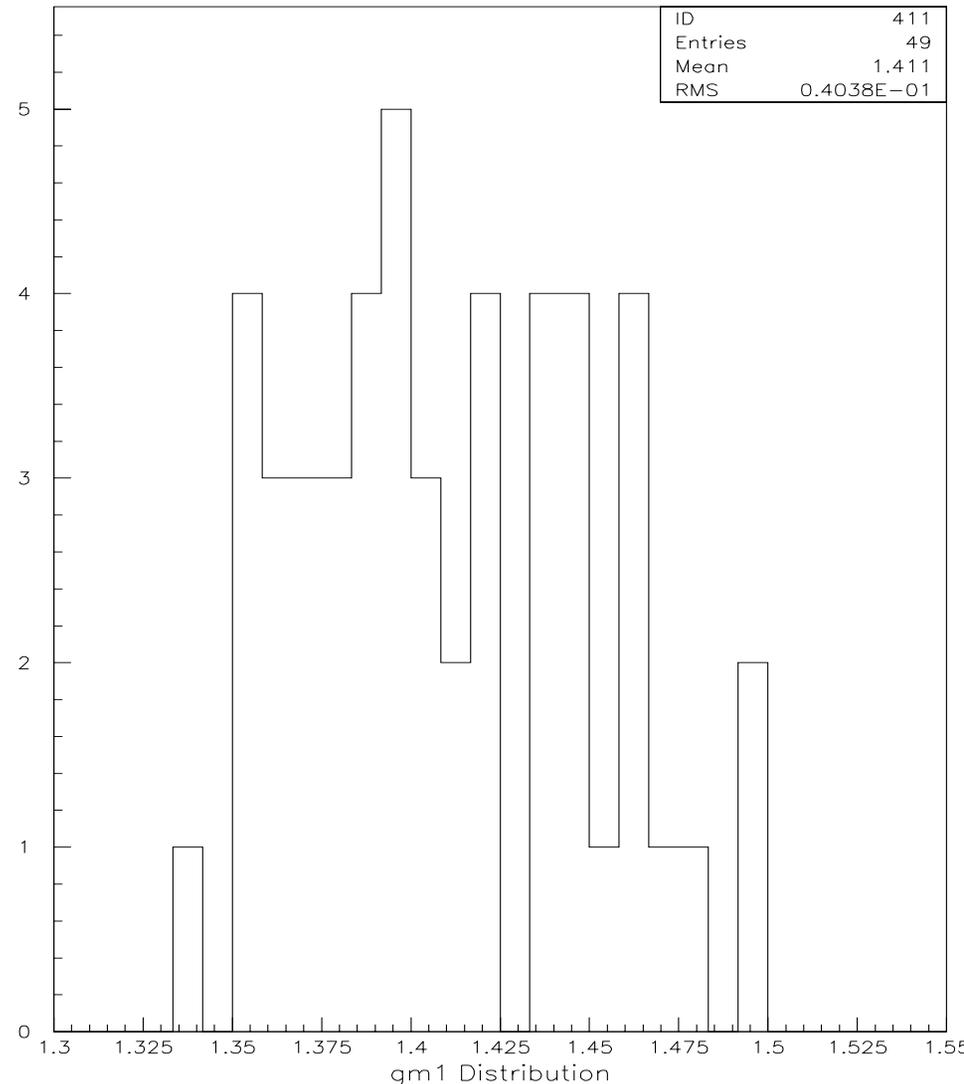
(\sim 0.5 hr per Q !!)

• Hence, currently:

• One Q=50 job takes \sim 1 day of CPU

• One Q=100 job takes \sim 2 days of CPU

→ Implies the need to carefully select a balance between a small amount of BMC error & the realities of CPU limitations



Theoretical Curve: $\sigma^2_{BMC}/\sigma^2_{DATA}$ vs. Q

• Assuming:

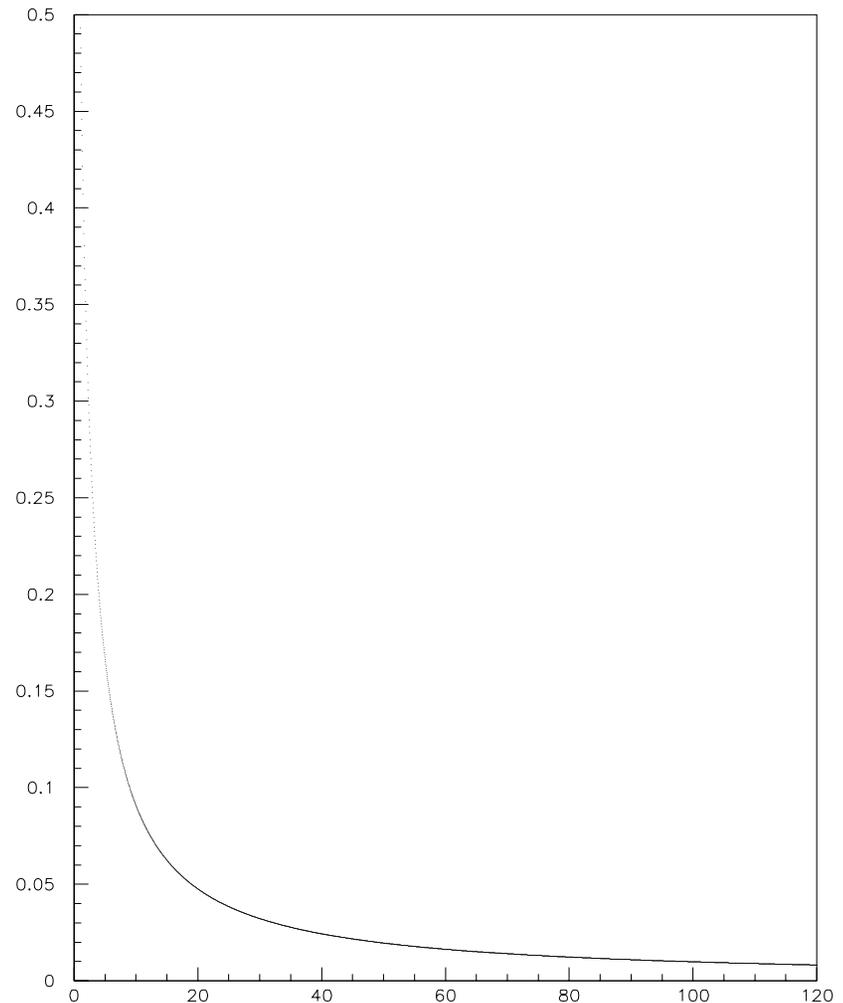
$$\bullet \sigma_{BMC}(Q) = \frac{\sigma_{BMC}(Q=1)}{\sqrt{Q}}$$

$$\bullet \sigma_{DATA}(Q) = \sigma_{BMC}(Q=1)$$

$$\bullet \sigma^2_{TOT}(Q) = \sigma^2_{BMC}(Q) + \sigma^2_{DATA}$$

• Then (with a little algebra):

$$\frac{\sigma^2_{BMC}}{\sigma^2_{TOT}} = \frac{1}{Q+1}$$



Current Projects (cont.)

- Getting all my v6_01 MC generation and analysis code running correctly on UVa's shiny new Linux cluster
 - Should provide ample CPU and disk space to generate sufficient MC to complete my statistical studies
- Optimization studies for analysis cuts

Upcoming Projects

- Complete statistical error analysis
 - (i.e. determine final values for σ_{BMC} and σ_{Ndata})
- Complete final background studies
 - Preliminary studies (with slightly different cuts) indicate that background contributions are small:
 - $K_L \rightarrow \pi^+ \pi^- \pi^0 \sim 0.3\%$
 - $K_L \rightarrow \pi^+ \pi^- \sim 0.1\%$
 - ** appears to be sufficient to account for virtually ALL observed background
- Extract fit parameter (g_{m1} , a_1/a_2 , g_{e1}) values from data, with statistical errors
 - Process is complicated by the large CPU requirements
 - Ultimately, approach may need to be slightly different than the one used for the $K_L \rightarrow \pi^+ \pi^- e^+ e^-$ analysis (which only has ~ 5000 events)
 - Final choice of “Q” for BMC will have major impact on time/difficulty of process
- Determine g_{e1} statistical sensitivity
 - (i.e. will my result be a value or an upper limit?)
- Begin study of systematic errors
 - Again: likelihood CPU requirements are likely to complicate this!